

**ESTIMATION OF FUTURE MANUFACTURING COSTS  
FOR NANOELECTRONICS TECHNOLOGY**

by

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## **(ABSTRACT)**

In this report, a future scenario concerning the economic direction of the computing industry has been presented. This future scenario was based on past developments within the computing industry. The continued miniaturization of semiconductor components was discussed based on observed trends for transistors. The physical limitations for transistor devices were also addressed. The use of x-ray lithography for the construction of devices on a “nano-scale” was considered. Next, cost trends within the microelectronics industry were explored. Although the cost per transistor has been observed to decrease, total equipment costs and facilities costs were observed to rise.

Trend extrapolation was next used to predict the future cost per transistor and the number of transistors per chip. By taking the product of these two predicted quantities, an equation for the future manufacturing cost per chip was determined. A parametric cost estimation model (VHSIC Model) for the prediction of avionics computer system costs was modified to reflect the future performance parameters of nanoelectronics. Using data from the x86 design of Intel® Microprocessor Chips, undetermined parameters of the Modified

VHSIC Model were calculated. Next, future performance parameters were used in the model to predict the initial selling price of future chips. The resulting predictions from this model indicated that chip prices are expected to increase while the price per electronic function will decrease. Finally, profit-time models for semiconductor chips and transistors were derived. These models were used to predict the future profit for a chip or transistor.

**Keywords:** Semiconductor economics, trend extrapolation, parametric cost models, nanoelectronics, nanotechnology, technology forecasting

## **DEDICATION**

To my dedicated mother and the memory of my departed father.  
I owe all of my success to them.

To my loyal and patient friends, most especially Aarti Khanna.  
All of your support is greatly appreciated.

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I also benefited from the assistance of other members of the MITRE Nanosystems Group. Daniel Mumzhiu assisted me with early library research. Group members Michael Montemerlo, Chris Love, and Johann Schleier-Smith shared their ideas and the research materials they had collected for a forthcoming MITRE review article on the technical aspects of next-generation, nanometer-scale electronics. I would like to thank the group for providing me with an unpublished manuscript of the forthcoming MITRE review article [Monte95]. This manuscript and other input from the Nanosystems Group [Nanos95] were used as a basis for the technical background featured in Chapter 1 and Appendix A of this report.

I would like to thank Rita Sallam and W. Allen Dogget, Members of the Technical Staff at MITRE. Their support at MITRE was helpful in the development of the ideas that form the backbone of this project.

The help from the Statistical Consulting Center at Virginia Tech was instrumental in the statistical analysis of the data for this report. Dr. Jerry Mann was helpful in the analysis of the regression models and Dr. Bob Foutz was insightful in the complete analysis of time-series data.

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# CHAPTER 1

## INTRODUCTION

### 1.1 Motivation

In today's computing environment, the transistor is the fundamental element of the microprocessor. As a device capable of processing much information quickly and efficiently, microprocessors have proliferated because of the adoption of desktop computers all over the world. The miniaturization of this fundamental component has proven to be among the most important development toward processing more information more quickly. In order to continue at the current rate of miniaturization, and to continue to increase the computing capability of electronic computers, fundamental new technologies for nano-meter scale electronics must be introduced. The subject of this report is an analysis of the economic prospects of these new technologies for nanoelectronics. In order to make these economic projections, it is first necessary to review some of the technical developments that provide the background for the introduction of nanoelectronics [Monte95, Nanos95].

Computer users and researchers continue to demand more power and the ability to process more complex information. To satisfy this demand in the next century, the microelectronics industry must explore fundamental changes in the design and manufacture of computing elements. This necessity has given birth to the field of nanoelectronics. Because nanoelectronics deals with the construction of transistors and electronic devices on a scale much smaller than current fabrication methods allow, new or modified manufacturing processes must be developed. Several fabrication processes have been proposed. However, the commercialization of such a process possibly lies up to 10 years in the future.

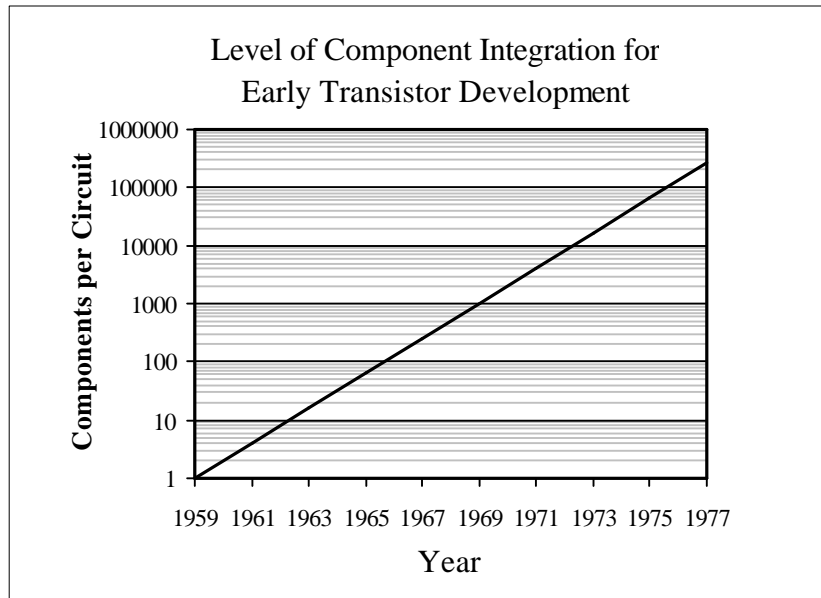
In this chapter, the ongoing efforts toward semiconductor device miniaturization will be summarized. The photolithographic fabrication process and the physical limitations to

continued miniaturization will be described. The development of x-ray lithography and electron-beam lithography as viable technologies for the future of electronics device fabrication will be described. The impact that the experimental field of nanoelectronics may have on the future of electronics computing will be highlighted. A more detailed description of the technical concepts presented in this chapter is included in Appendix A. The problem statement and the scope and limitations for this project will be stated. The plan of presentation for this report will conclude the chapter.

## **1.2 Development of Semiconductor Computing**

The “state-of-the-art” in the late 1940s before the advent of solid-state transistor devices was the renowned ENIAC computer developed at the University of Pennsylvania. ENIAC contained 19,000 vacuum tubes and was housed in a very large room that needed constant cooling because of the large power consumption of the tubes [Noyce77]. Subsequently, vacuum tubes were replaced by solid-state devices based on the transistor. This not only improved the reliability of electronic computers, but it also substantially reduced the power required and associated heating problems. However, even more dramatic improvements in computer performance appeared as a result of the miniaturization of such solid-state electronic devices over the next 45 years. The increase in the density of transistors within electronic devices is plotted versus time for a portion of that period in Figure 1-1.

In the late 1970s the trend slowed to roughly a four-fold increase in transistor density every three years. This trend, known as Moore’s Law [Hutch96] has been observed to continue from 1977 through 1996.



**Figure 1-1:** The level of component integration for early transistor development is shown. Increase in component integration was observed to double every year from 1959 to 1977. (Adapted from Noyce, R. N., “Microelectronics,” Scientific American, Sept. 1977.)

### 1.3 Current Silicon Semiconductor Technology

Current silicon semiconductor technology is very effective in producing mass quantities of high-performance computer chips. Following is a brief description of the fabrication process. Further reductions in feature size are becoming increasingly difficult because the physical limitations of the fabrication process can no longer resolve significantly smaller dimensions. Furthermore, current technologies for preserving the photo-lithography process while decreasing the minimum feature size of electronic circuits are discussed.

#### 1.3.1 Fabrication Process

Photolithography occurs over a series of steps where thin films of various photo-sensitive materials (photoresist) are exposed to ultra-violet light on the surface of a pure silicon substrate roughly 20 centimeters in diameter. Removal of desired section of photoresist exposes the substrate to etchants that transfer the circuit pattern to the surface of

the wafer. Doping of the wafer exposes the substrate to impurities such as boron or arsenic. These impurities alter the electrical conductivity of circuit elements that are components of the transistors that comprise the microprocessor [Stix95]. The finished wafer is then cut up into individual chips that can number up to 200 for the 20 centimeter wafer [Hutch96]. A pictorial description of the chip fabrication process is shown in Appendix A of this report.

### **1.3.2 Physical Limitations**

Transistors fabricated using photolithography have been remarkably easy to scale down. As minimum feature sizes for transistors continue to decrease, their operational characteristics will become affected. In addition, the manufacturing technology necessary to obtain such small dimensions will become more complex, and obviously more expensive.

Current manufacturing technology has the capability to produce transistors with a minimum feature size of 0.35 microns (350 nanometers) [SIA94]. The ability to continue to produce devices with smaller feature sizes is dependent on various factors in the manufacturing equipment and physical limitations of the transistor elements. Some of these factors are listed below [Stix95].

1. Wavelength of Emitted Light
2. Lens Focus on Photoresist
3. Heat Dissipation
4. Tunneling

A full description of these limitations is included in Appendix A of this report.

### **1.3.3 Lithography Solutions**

With the limitations that are being slowly realized in the semiconductor industry, manufacturers are attempting to circumvent those limitations while trying to maintain the lithographic technologies of the past. Two of the more prominent proposals are the use of x-

ray lithography and electron beam lithography. A brief description for both of these technologies as well as their advantages and disadvantages are presented below.

#### *1.3.3.1 X-Ray Lithography*

Fine resolution images have been achieved through the use of x-ray lithography. At roughly one nanometer, the wavelength of x-rays are about three hundred times smaller than the light used in today's commercial systems. While conventional lithography systems can use the radiation emitted from advanced lasers for production, x-ray lithography equipment must use the radiation generated from a synchrotron as a source of radiation [Stix95]. A synchrotron is an energy source that consists of two superconducting magnets whose electric field confines electrons within a closed orbit. The electrons that circulate within the storage ring emit x-rays that are used in the lithography process [Stix95].

Although synchrotrons are available in only a few of the most advanced universities, IBM has developed the only commercial synchrotron storage ring in the United States [Stix95]. IBM's synchrotron system still remains a developmental project. A primary technical obstacle to the commercial development of x-ray lithography is the lack of a feasible way to focus x-rays. Also, x-ray lithography lacks the ability to demagnify the image. As a result, the masks must outline the circuit image at the same small size as the image created on the chip. The development of masks that can absorb the generated x-rays also poses a problem. Since x-rays are high energy electromagnetic waves, masks must be made very thick in relation to circuit features so that x-rays will not penetrate the mask [Stix95]. For successful development of x-ray lithography the technical obstacles mentioned above must be overcome.

#### *1.3.3.2 Electron Beam Lithography*

The use of electron beam lithography has a tremendous capability to produce high-resolution features on silicon wafers. Resolution as small as tens of nanometers can be

achieved as a result of two factors: (1) electron beams have much shorter wavelengths than current photolithographic light sources and (2) the electric fields that control the beams can be focused very accurately [Gent94]. Electron beam lithography writes like a fine writing instrument since the focus of the beam is computer-controlled. Slow production rates, in comparison to conventional photolithographic techniques, make electron beam lithography both time and cost inhibitive [Gent94].

#### **1.4 Nanoelectronics: The Future of Computing Technology**

The term nanotechnology has been used to describe a field of research that deals with the manipulation of matter on an atomic scale. A nanometer, one-billionth of a meter ( $10^9$  meters), is only about 10 atomic diameters. The fabrication of a computer based on nano-scale components would realize a density of up to 10,000 times more than today's most advanced computers [Monte95].

The large microelectronics infrastructure currently in place will be an advantage for the emergence of nano-scale computers based on electronic characteristics. Considering the obstacles to continued miniaturization within semiconductor manufacturing technologies, the operating principles that govern the devices within these "nanocomputers" must take advantage of the limitations that are currently being experienced [Hans91]. Fabrication techniques will eventually determine how small the devices can be built in practice.

The proposed nanoelectronics solutions to the obstacles of the microelectronics industry come from diverse fields of interest. Researchers from the fields of biology, chemistry, physics, and mechanics have proposed computational devices from their respective fields [Monte95]. The true test of these systems will ultimately be whether their enhanced performance characteristics will justify increased manufacturing costs when compared to today's most advanced computing systems.

Because smaller device sizes will allow semiconductor manufacturers to place a larger number of logic gates within a chip, more computational power will be achieved. The application of nanoelectronics manufacturing will also be applied to memory systems. The development of nano-scale devices will not only provide the computer user with more power and information storage capacity, but it may benefit society in the solution of previously insoluble problems [Rola91]. For more information on the technologies, devices and fabrication techniques for nanoelectronics see Appendix A for a detailed explanation.

## **1.5 Problem Statement**

The problem under investigation is the estimation of selling prices and manufacturing costs for nanoelectronics technology. Projections of current trends will indicate the future manufacturing costs of semiconductor chips. Price estimates will be calculated by modifying technical parameters of previous cost models for microelectronics. From these two projections, the profit per chip and transistor can be obtained.

## **1.6 Scope and Limitations**

Despite the comprehensive description of this research, there are several limitations. This report estimates the future manufacturing costs of nanoelectronics technology based on trend extrapolation and model modification. Because of the proprietary nature of the models that some industry experts use, those models were not available to be included in this research.

Only the silicon-based semiconductor electronics industry is examined in this report. It is realized that the development of nano-computing may emerge within fields other than electronics. Much effort has been expended to create nano-computing systems in quantum, mechanical, and chemical fields. A more comprehensive description of these fields is included in Appendix A of this report.

In addition, this report has attempted to quantify only the manufacturing costs and selling prices for the future of nanoelectronics manufacture. The quantification of the enhanced performance characteristics for nanoelectronics is not addressed.

## 1.7 Plan of Presentation

| <b>Chapter</b> | <b>Description</b> |
|----------------|--------------------|
|----------------|--------------------|

- |        |  |
|--------|--|
| One:   | introduces the continued efforts of product miniaturization within the microelectronics industry. Limitations and remedies for the photolithography process are described. The developmental technologies for the future of electronics device fabrication are described. Limitations for the context of this report are explained.  |
| Two:   | explains the cost trends and capital expenses for current microelectronics manufacturing. Two examples of cost models for the manufacture of silicon-based semiconductor devices are described and analyzed. Based on analogies from past technologies, technology forecasting is used to describe the hurdles for nanoelectronics development.  |
| Three: | provides an analysis of the two described cost models (cost trend extrapolation and the VHSIC parametric cost model). Modification of the VHSIC model is performed based on the model's limitations, information available, and the model's applicability to nanoelectronics. Performance factors such as chip processing frequency and number of components on the chip are included in the Modified VHSIC model. Time-based extrapolation of trend data is performed to predict future chip costs. |
| Four:  | estimates the initial offering price of future chips based on the modifications made to the VHSIC cost model. The future estimates for initial chip offering price are obtained by first determining the unknown parameter values in the modified model. Substitution of future performance parameters is performed so that a future estimate of initial chip offering price can be obtained.  |
| Five:  | analyzes the projected costs for nanoelectronics manufacturing. A profit-time equation is derived for the estimation of future profits on a per-chip and per-transistor basis. Also, the enhanced computational characteristics as a result of improved device design are considered.  |
| Six:   | provides quantitative conclusions, describes the utility of this report as a resource for industry to guide capital investment decisions and makes several recommendations for future research.  |

## **CHAPTER 2**

### **LITERATURE REVIEW**

In this chapter, the cost trends for current microelectronics manufacturing will be outlined. These cost trends will describe some of the unit production costs as well as the facilities costs for the production of microelectronics circuits. The available cost data for the development of x-ray lithography will also be included. A general form of a parametric cost model will be given. Finally, cost trends and models for silicon-based semiconductor manufacture will be described and analyzed.

Technology forecasting will be examined as a framework for the estimation of the impact of innovative technologies. The impact of new manufacturing technologies for electronics will be examined from analogies based on past technologies. By using this approach, a prediction can be made for the future performance characteristics of electronics. The limitations and cost factors for the previous technology will be discussed and analyzed within the context of technology forecasting.

#### **2.1 Microelectronics Cost Trends**

One of the primary advantages for the development and production of microelectronics circuits is the ability for those circuits to be manufactured in large batches at low cost. Advances in manufacturing technology have allowed for the production of microelectronics circuits with smaller features. These smaller features have allowed a greater density of components to be placed on a chip and have reduced the cost per component within these chips. The cost per component has decreased primarily because the technology for placing more components on a chip has advanced more quickly than the capital costs for such equipment. As manufacturers face technological limits for the processing of microelectronics circuits, the ability to produce these circuits at such low unit costs may be weakened. This

section of Chapter 2 will describe some of the cost considerations for the microelectronics industry. Some of the unit cost trends will be discussed as well as the capital investments for the facilities will be addressed.

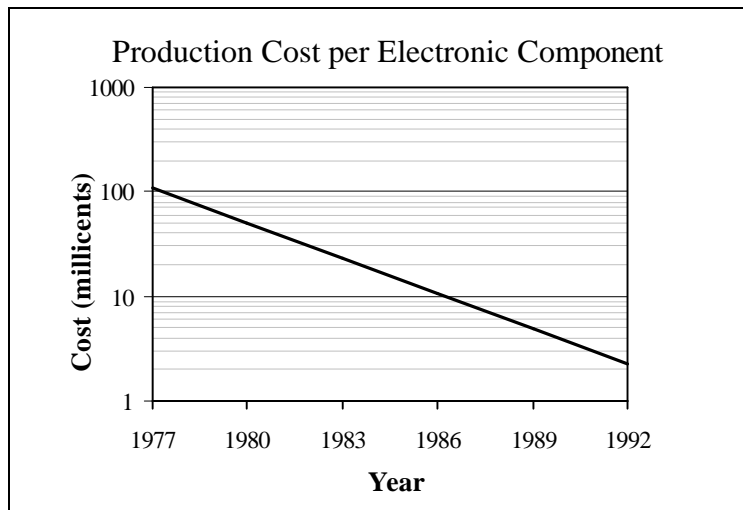
### **2.1.1 Unit Cost Trends**

The cost trends that will be addressed in this section primarily deal with the cost per electronic function or cost per electronic component within a microelectronics circuit chip. Such costs are usually computed by dividing the manufacturing cost per chip by the number of components (transistors) within the chip. As shown in Figure 1-1, the number of transistors on a chip has been shown to exponentially increase since the advent of solid-state transistor electronics in the late 1950s. This exponential increase for the number of transistors within a device has been met with an exponential decrease in the cost per transistor [Hess92]. This is not to say that the costs for the manufacturing technologies used to produce these circuits has decreased. The capital investments for these technologies have increased, however, technology advancements have allowed manufacturers to place more transistors on a computer chip at a lower cost per transistor.

One example of the cost trends discussed in the previous paragraph is shown pictorially in Figure 2-1. The production cost per electronic component for the production year of a microelectronics chip is shown. The data for this figure was adapted from the 1994 National Technology Roadmap for Semiconductors. Although the Roadmap included production costs for several areas of semiconductor manufacture (logic transistors and memory transistors), the production costs shown are for the logic transistors used in microprocessor chips [SIA94].

This figure clearly shows that there is an exponential relationship for the production cost of transistors and the year in which the transistor was produced. As stated earlier, the primary reason for this significant decrease in cost is that technology advances allow

manufacturers to increase transistor density at a faster rate than the increases in capital equipment costs. Only the production costs for solid-state production techniques are included in this figure. No significant deviations for the unit cost trends were observed throughout the years included in this figure.



**Figure 2-1:** The production cost per electronic component in a microelectronics chip is shown. Production cost was observed to half every three years. Adapted from SIA, “The National Technology Roadmap for Semiconductors,” 1994.)

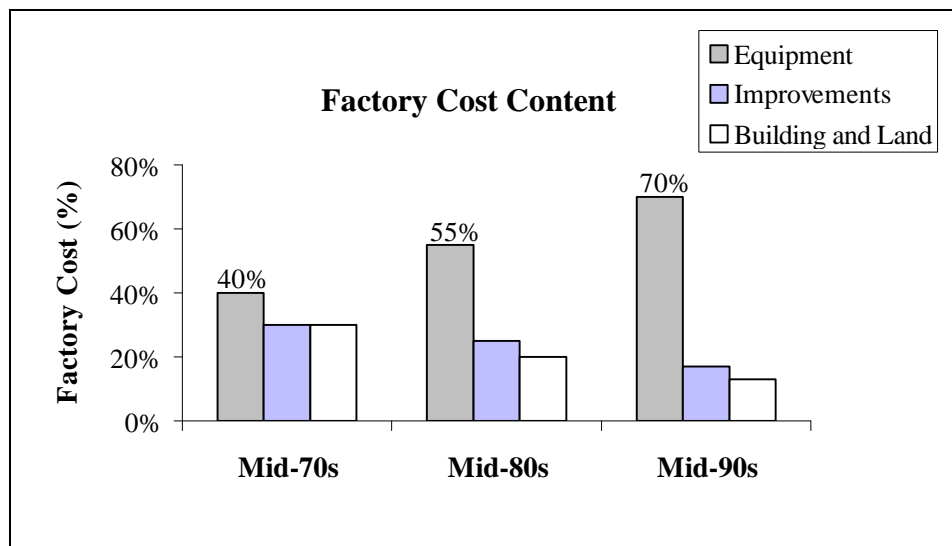
### 2.1.2 Production Costs

Despite the lowered costs for the production of transistors on a unit basis, the actual production costs per chip (and wafer) have been observed to increase with each generation of processing equipment. These increases are described in the following section. The increasing contribution of equipment for the Factory Cost Content is described. Factory Cost Content is the contribution of resources (equipment, building, land, and improvement) to the total production cost. Costs for the equipment similar to the ones used in today’s fabrication facilities are listed in this section. As well, the estimated overall facilities costs are included for some of the leading semiconductor manufacturers.

### 2.1.2.1 Equipment and Factory Cost Content

As the equipment used in microelectronics factories becomes more complex and expensive, the equipment's contribution to the total cost of the factory is rising. With each successive generation of equipment, the resolution for microelectronics chip production becomes more precise. This resolution is largely dependent on the overlay of the processing equipment. Overlay defines the precision with which an image can be placed on the chip [SIA94]. Overlay is the determining factor in the complexity and accuracy for the development of more advanced manufacturing equipment. As such, the cost for each successive generation is affected by that complexity.

The contribution of the factory cost as a result of equipment expenditures can be seen in Figure 2-2. This figure shows the percentage of Factory Cost Content from three categories: Equipment, Improvements, and Building and Land. As shown in this figure, the contribution that equipment has to the Factory Cost Content was observed to rise from 40% in the Mid-1970s to approximately 70% in the Mid-1990s.



**Figure 2-2:** The Factory Cost Content is shown for three categories: Equipment, Improvements, and Building and Land. In the past two decades, contributions from equipment have been observed to rise sharply [SIA94].

### 2.1.2.2 *Equipment Costs*

Figure 2-2 shows that a critical cost component for semiconductor manufacturers is equipment. The largest component of this cost is photolithography equipment (steppers). Two important considerations for any long-term ownership are the initial cost of the stepper and the long-term costs of ownership [Rub90]. The initial cost of the stepper sometimes dictates the decision to purchase one stepper system over another. However, the costs of ownership can often dictate the contribution that the stepper system will have to the overall manufacturing cost of the product.

Several primary factors in the cost of ownership for photolithography equipment are the annual depreciation, processing materials (resists and masks), the space required within a cleanroom, and the spare parts requirements as a result of machine reliability. As stepper systems become more advanced, both the initial investment cost and the cost of ownership are expected to rise. Some of the initial investment and ownership costs are included for several photolithography stepper systems in Table 2-1. The data for the following table were taken from an article by Daniel Rubin, that appeared in *Microelectronic Manufacturing and Testing* [Rub90].

**Table 2-1: Initial Investment and Ownership Costs for Photolithography Systems**

| <b>Lithography Equipment</b> | <b>1X</b> | <b>5X</b>  | <b>5X Stepper</b> |
|------------------------------|-----------|------------|-------------------|
| List Price                   | \$750,000 | \$1,500,00 | \$1,800,000       |
| Wafer Production/Year        | 19,106    | 18,876     | 23,348            |
| Direct Labor/Wafer           | \$17.99   | \$26.40    | \$26.40           |
| Stepper Cost/Wafer           | \$43.92   | \$71.30    | \$69.87           |

An important observation from the data given above is that although the initial investment cost for the new 5X stepper is \$300,000 higher than the traditional 5X stepper, the new stepper has a higher throughput. This additional investment should be considered for

those manufacturers who desire a high throughput of similar or identical products. The higher throughput may justify the additional investment in the new stepper for some manufacturers.

Continued technological progress in photolithography has shifted the light used in the fabrication process to lower and lower wavelengths. The photolithography equipment described above exposes wafers in the visible portion of the electromagnetic spectrum (approximately 380 nm). Low wavelength ultra-violet light (still in the visible portion of the spectrum) is still being used to expose wafers. Preliminary research for the use of x-rays to expose wafers has been undertaken by IBM. X-Ray sources emit electromagnetic waves that are not in the visible portion of the electromagnetic spectrum (approximately 10 to 30 nm). Many researchers feel that the research for x-ray lithography techniques was performed much too early [Stix95]. However, processing limitations in the visible portion of the spectrum are becoming more evident with each new generation of conventional lithography equipment. Following is a description of the costs incurred by IBM from their preliminary research of x-ray lithography.

According to Stix [Stix95], IBM has performed initial research for the use of x-rays as an exposure source for the processing of silicon wafers. This research primarily consisted of designing a synchrotron to produce x-rays that could supply multiple steppers with the 'light' to expose the wafers [Stix95]. The price tag for the development of a synchrotron was estimated at \$20 to \$50 million. An advantage for such a system is that a single synchrotron could supply up to 16 steppers simultaneously. At present, each stepper has an independent exposure source. Other technical obstacles that prevent IBM from fully developing this x-ray system include: focusing the x-rays, reduction of the image, fabrication of the masks and development of resists. In total, IBM spent several hundred million dollars on the research [Stix95]. Stix mentions that in late 1995 an alliance within the semiconductor industry would decide whether such a system should be prepared for production. Information regarding any decision from the alliance has not been identified.

### 2.1.2.3 *Facilities Costs*

Another key piece of evidence that the technological and economic limits to semiconductor manufacture are close is the sharp rise in facility costs. The primary cost driver for the sharp rise in facility costs is the clean rooms within which processing of the silicon wafers occurs. Processing rooms must be nearly particle-free. The clean room requirements for such facilities drive initial costs to over five thousand dollars (\$5,000) per square foot [Cast94]. This price only includes the construction cost, the specialty equipment cost, and hookup costs for the equipment. Excluded from this estimate are the costs for the actual processing equipment used in the wafer fabrication processes [Cast94].

Despite these large costs, semiconductor manufacturers continue to heavily invest in the construction of these facilities. Intel is spending \$1.1 billion for a factory in Oregon and \$1.3 billion for a facility in Chandler, Arizona. Samsung and Siemens are building plants that are each estimated to cost \$1.5 billion. Motorola, IBM, and Toshiba also have plans for the construction of facilities costing over one billion dollars each [Hutch96]. A direct consequence of the rise in facility costs is the formation of alliances between the semiconductor manufacturers to share technological advances and development costs. Such alliances will be instrumental in the development of a commercial x-ray lithography system.

## **2.2 Cost Models**

For a system that has yet to be developed, the generation and analysis of a cost model is not only desired, but necessary to justify the development costs for production. As new manufacturing processes for microelectronics are developed, the economic success of such systems are determined by their cost effectiveness and the benefit generated for a particular cost. The economic benefit is directly measured by the market selling price of the products developed by the particular system. Systems whose economic benefit exceeds the costs generated by that system are economically desired provided that the margin of excess benefit exceeds an expected hurdle or percentage rate. Accurate cost models not only help to predict

the costs associated with a particular system, they help to identify the economic benefit for the system or product being developed.

Cost models for the microelectronics industry have been used for a variety of purposes. Their implementation for the determination of facilities' costs, ownership costs for equipment, and cost trends associated with the cost per transistor of microelectronic devices are well known. As new manufacturing technologies are developed, these cost models must be modified or new cost models must be generated. These new or modified cost models must reflect the changes or advances that have occurred since the previous cost model was developed. In this section, a generalized parametric cost model is described. A parametric cost model can be used to predict system costs based on technical parameters relevant to the system being investigated. Also, two specific cost models for microelectronics production are described. The first estimates production costs based on technical parameters. The second model assumes that the current trends in microelectronics development will continue in the future.

### **2.2.1 Dean's Generalized Cost Model**

Edwin Dean [Dean89], has described a generalized parametric cost model used for the estimation of cost during system design. This generalized cost model uses technical metrics for the estimation of system costs. The metrics are equations called cost estimating relationships (CER's) and are obtained by the analysis of cost and the technical parameters to be included in the model [Dean89]. A critical assumption that Dean makes in this formulation is that a measurable relationship exists between the technical parameters and the costs to be estimated [Dean89].

Dean states that "candidates for metrics include system requirement metrics and engineering process metrics." [Dean89] The resulting system cost is reflected in a model that

provides a mathematical relationship to the system cost being estimated. Dean's Generalized Cost Model is stated below:

$$c = e^{a_0} \prod_{i=1}^r e^{a_i x_i} \prod_{j=r+1}^s x_j^{a_j}$$

Dean states that the exponential factors of the model are useful for normalizing cost of temporal effects of technology escalation and binary categories [Dean89]. The power law factors are useful for normalizing cost for economic quantities of production units. This generalized model is advantageous over simple models because the "combination of these factors usually improves accuracy over the use of either separately." [Dean89]

### 2.2.2 Advanced Technology Models for VHSIC

Efforts to determine costs for Air Force Avionics have produced a cost model that estimates the costs for the implementation of VHSIC (Very High-Speed Integrated Circuits) in avionics systems. VHSIC was initiated by the Defense Department to skip an entire generation of microelectronic technology and move military electronics into the era of commercial and consumer electronics [Hess92]. This analysis is presented in a report entitled: "Estimating the Cost of Advanced Technology," by the RAND Corporation. In the report, the cost implications of VHSIC are discussed. Also, an approach to performing cost estimates for such systems is presented. Although the cost estimates performed by RAND do not directly deal with cost estimation for the fabrication of microelectronics circuits, the methods applied for estimation of cost are of primary interest.

The VHSIC Cost Estimation Model specifically addresses the costs for the development of avionics computer systems. The VHSIC Model is similar to the generalized model described in Dean's report. This similarity shows that the RAND Corporation used an standard and accepted format for the development of their cost model. The challenge for the

development of these estimates is that the development of such a system is unlike any systems that appeared in historical cost databases. The significance of making such cost estimates was that the electronics within avionics systems was found to contribute up to 50% of the total cost of the aircraft [Hess92]. Therefore, accurate cost estimates were desired early in the design/manufacturing process.

A technology forecasting approach was used to estimate costs in the VHSIC model. The cost trends described earlier in Figures 1-1 and 2-1 were used to approximate a relationship between the level of integration on the circuit chip and the cost per component on the chip. The authors asserted that if such a relationship could be projected, it could be used to help forecast the cost of future systems [Hess92]. For the specific application developed by the RAND Corporation, the researchers chose to select specific cost drivers that reflect the established cost trends for microelectronics development. After considering a number of candidates to be included in the theoretical cost model, three cost drivers were selected. These three drivers are the number of gates on the chip, the weight for the system, and a binary variable to reflect the maturity of the technology being utilized [Hess92]. This binary variable was chosen to reflect the utilization of a product before or after two years from its initial availability. If the technology was implemented during the first two years of its availability, the variable is given a value of 1, otherwise the value is 0.

The researchers at the RAND Corporation expected the relationships between the parameters specified within the model and the system cost to be non-linear. As such, the VHSIC model was observed to be logarithmic-linear. The model used by the RAND Corporation is shown in Equation 2-1. The reader should note that the model shown below is similar to the general parametric cost model developed by Edwin Dean.

$$T_{100} = b_0(lbs)^{b1} (LOI)^{b2} e^{b3*(NEW)} \quad \text{Equation 2-1}$$

where:  $T_{100}$  = Unit 100 production cost in thousands of 1987\$

- LBS = System weight (in pounds)
- LOI = Level of integration (number of transistors/chip)
- NEW = 0 mature technology, 1 new technology
- $b_i$  = numerical constants (no units)

Since RAND expected costs to increase at a decreasing rate with weight and level of integration, the values of  $b_1$  and  $b_2$  were specified to be less than one.

For the computer subsystems used in aircraft avionics, the cost estimation model developed by the RAND Corporation is shown below in Equation 2-2. The value of  $r^2$  for this relationship was determined by the RAND Corporation to be 0.88. The value of  $r$  is known as the correlation coefficient and indicates the strength of the relationship between the variables. A value close to unity is most desired for the correlation coefficient. The standard error of the estimate was 0.37.

$$T_{100} = 2.20(lbs)^{0.95} (LOI)^{0.07} e^{0.80*(NEW)} \quad \text{Equation 2-2}$$

In their conclusions, the researchers from RAND stated that reliable cost estimates for the future of combat aircraft avionics systems were difficult to obtain. The authors also stated that the application of parametric methods were difficult to apply because so little historical cost experience was available.

### 2.2.3 Estimates Based on Cost Trends

The cost trends for the development and production of microelectronics circuits are well known and documented in the literature. One of the most common ways in which the costs of microelectronics components can be represented is that the cost per electronic component is shown throughout a specified time period. This type of representation usually shows a sharp decrease in cost per component as time and technology advances. Such a relationship is shown in Figure 2-1 of this report. An exponential decrease in cost is observed

primarily because technology advances have been implemented faster than increases in processing costs.

The cost trends represented can exist for any component of the microelectronics fabrication. In Figure 2-1, the trends are represented for logic transistors fabricated for microprocessors. These microprocessors are typically used in desktop computer applications. The trends developed in Figure 2-1 were adapted from data given by the Semiconductor Industry Association (SIA). This data showed the historical relationship between the cost per transistor and the year of development for the microprocessor. The data are given in Table 2-2, below:

**Table 2-2: Logic Transistor Cost from 1977 to 1992**

| <b>Year</b> | <b>Cost (millicents)</b> |
|-------------|--------------------------|
| 1977        | 112                      |
| 1980        | 47                       |
| 1983        | 24                       |
| 1986        | 13.5                     |
| 1989        | 5.2                      |
| 1992        | 2.7                      |

Considering the cost trends that have occurred in the past, cost projections can be made for the future component transistor costs of microelectronics. These cost trends can be determined by doing a least-squares linear regression of the data presented below. The linear regression would be performed on the year and the antilog (base 10) of the data provided in Table 2-2. By determining a logarithmic-linear equation of the data, the unit transistor costs could be estimated for any year of production. The primary assumption of this type of approach would be that no significant deviations from the observed trends would occur.

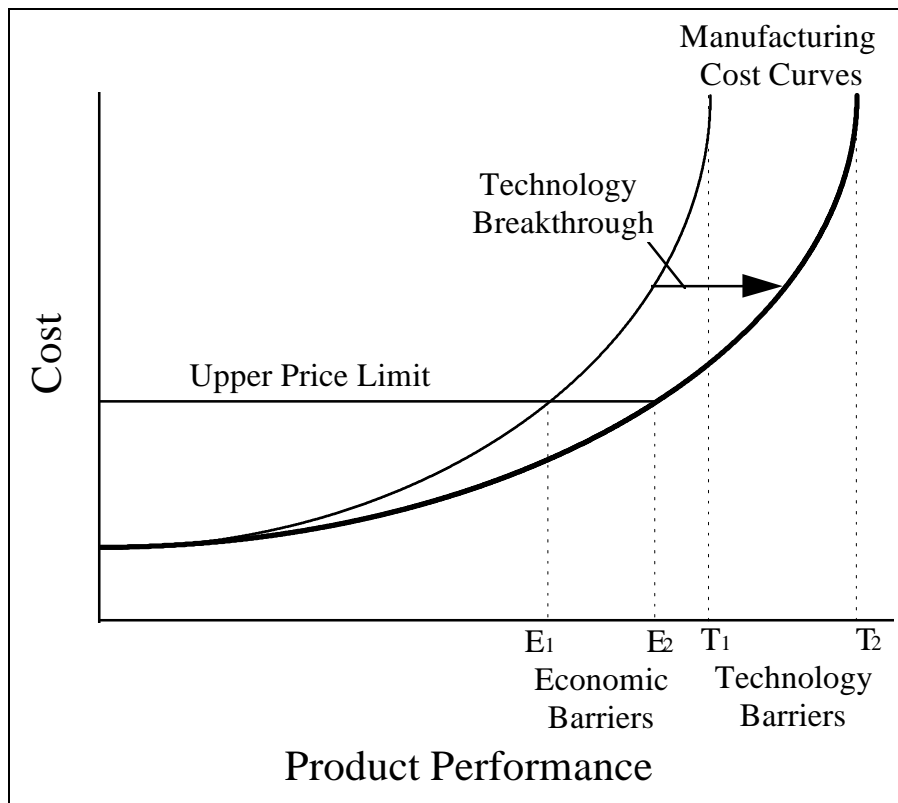
### 2.3 Technology Forecasting

Technology forecasting describes a field that is interested in the logical analysis of systems that leads to a performance objective or hurdles for the development and implementation of a particular technology. Such conclusions assist and identify the economic potentials and impact of technological progress [Lan69]. This field differs from opinion-based conclusions in that technological forecasts rest upon an explicit set of quantitative relationships and assumptions [Brig68]. Three primary groups that are interested in the outcome of these forecasts are business economists, managers of corporate technology, and government agencies such as the military. Such organizations may use these forecasts to predict future business trends, to decide future product strategy, or to determine future systems capabilities [Brig68].

A common approach for the implementation of technological forecasts is the observation of performance or cost trends within the industry of interest. By identifying the trends that characterize the industry, extrapolation of those trends can predict the future performance characteristics for the parameter of interest. The use of technology forecasting through the use of trend extrapolation avoids the problem of making detailed predictions about the development of specific devices. A disadvantage, however, is that trend extrapolation merely states that a certain level will be achieved. It does not indicate whether existing devices can be improved or whether a new device will be invented [Mart72]. By using several parameters to identify the advancement of the industry, the performance or cost of a future system can be predicted. Such a method is called parametric cost estimating based on technological forecasting.

As technological forecasting can predict the enhanced performance of a new or revolutionary technology, some attention must be given to the effects of technology breakthroughs. Figure 2-3 shows the effect of technological progress on the manufacturing cost of a product. The technological barriers  $T_1$  and  $T_2$  are where improvements in product

performance are achieved only at a very high cost. In this example, economic barriers are encountered before technological barriers [Hutch96]. Technological breakthroughs, as represented by the arrow, have the effect of bending manufacturing cost curves downward. This effect allows the manufacturer to provide a product with enhanced features at the same cost or the same features at a lower cost.



**Figure 2-3:** The contribution of technology breakthroughs on the overall system cost is shown. Note that as technological barriers are overcome, enhanced product performance is obtained for the same cost [Hutch96].

### 2.3.1 Aircraft Technology Forecast

A field that generated many forecasts for future performance characteristics is the airline industry. The rapid development of this industry led many to speculate about the possibilities of shifting from propeller-driven aircraft to jet engine airplanes. Both the aircraft

and semiconductor industries were originally driven by military demand for their product [Hutch96]. For aviation, the shift from military application to a consumer product was driven by the reduction in cost per mile traveled, while also reducing the flight time. For semiconductor electronics, consumers have benefited from the efforts of industry to increase component densities (also increasing performance), while lowering chip costs [Hutch96].

A primary indicator of performance in the aviation industry is maximum speed. As aviation development grew, the speculation for advanced performance led some industry experts to predict supersonic travel. These predictions were later verified with the invention of the Concorde. This invention, and its application for supersonic trans-oceanic travel, is still a subject of much debate. Although the increase in speed was a natural extension of past trends, Concorde travel is still not economically feasible for most travelers. A major pitfall for the adoption of widespread use of Concorde travel was the noise pollution generated from the sonic boom of deceleration [Hutch96]. Despite the limitations for advancement, the airline industry did not falter. Instead, a greater diversity of smaller aircraft were designed and built for more specific markets [Hutch96]. The focus of development shifted from size and speed to efficiency, quality, and comfort [Hutch96].

### **2.3.2 Hurdles for Nanoelectronics**

Similar to the development of the Concorde, nanoelectronics is expected to be a natural extension of an observed trend. That trend, the level of integration for components on a microprocessor chip, has been observed to increase exponentially for the past three decades. Although the obstacles for continued development appear insurmountable, semiconductor industry researchers have previously overcome similar obstacles [Stix95]. The continued demand for semiconductor electronics will fuel the research for more innovative solutions for the ongoing miniaturization of the transistor.

Among the obstacles that must be overcome, the most prominent is the search for a lithography system that can resolve smaller images [Stix95]. For much smaller wavelengths, the application of x-ray lithography appears to be an alternative with several major hurdles for full commercial development [Stix95]. Also necessary is a circuit design that will take advantage of the quantum effects experienced for such small circuit features [Bate88]. The immediate solution to these obstacles is not clear. However, technology forecasting can be used as a tool to determine the future performance level of semiconductor devices.

## **2.4 Summary**

This chapter has described some preliminary considerations for cost trends associated with the continued development of the semiconductor industry. One of the primary advantages for the continued decrease in semiconductor costs is that technological advancements have outpaced the increased cost of manufacturing computer chips. However, increases in the total cost of processing equipment have been observed to rise over the past two decades. As manufacturers seek to develop more advanced manufacturing hardware, the facilities costs have also risen. Facilities currently under construction may cost in excess of one billion dollars. The primary contribution for the large cost of these facilities is the equipment for processing and the preparation of the clean-room environments.

Two methods for the estimation of future costs have also been discussed in Chapter 2. These costing methods use industry trends to extrapolate a future industry parameter. As observed from the generation of the VHSIC Cost Model, the application of technology forecasting is possible for the prediction of future performance parameters. These parameters can then be used in a cost model to estimate the future manufacturing costs of an industry. A similar study is to be performed for the commercialization of x-ray lithography manufacturing systems.

## **CHAPTER 3**

### **MODIFICATION OF COST MODELS**

The VHSIC Model and the cost trends described in the previous chapter were shown to address some of the past and current production costs for microelectronics. Although the models discussed may accurately describe the manufacturing costs for past or current microelectronics systems, their application to future systems is limited. The primary purpose of this report is to estimate the future manufacturing costs of nanoelectronics. Therefore, the cost models previously identified must be modified to reflect the future properties or performance characteristics of nanoelectronics.

In this chapter, the advantages, disadvantages and complexity of the previously described cost models are addressed. Considering the technology forecasting analogy, predictions for future performance factors will be made for nanoelectronics. These predictions will be extrapolations of currently observed industry trends. It is realized that the cost models analyzed may not reflect future manufacturing costs based on the specified performance characteristics or the available data. In that case, the models will be modified to reflect the performance characteristics or manufacturing technology of the next generation of electronics. Such performance characteristics include, but are not limited to number of components on the chip, product yield, minimum feature size of the devices on the chip, and processing frequency of the integrated circuit chip.

#### **3.1 Analysis of Cost Models**

The primary advantage for the development of a cost model is to obtain a macroscopic view of costs by using several measurable parameters. These parameters should exist within the system or process that is being examined. The two models discussed in Chapter 2 are a

parameter-based model for the estimation of microelectronics costs and a cost trend model for the estimation of per transistor costs. In this section of Chapter 3, the advantages, disadvantages, and complexity of each of these models is discussed. The relevance and applicability of these models is also addressed. The generic model specified by Dean will be used as a basis for the development of a parametric-price model for nanoelectronics. Portions of the VHSIC model will be used to generate a Modified VHSIC Model. Later, chip costs will be estimated by extrapolating chip transistor content and transistor costs.

### **3.1.1 VHSIC Model**

The VHSIC Model developed by the RAND Corporation for cost estimation of avionics computer systems is identical in form to the general model specified by Dean. Although future nanoelectronics technology will not be identical to the computer system technology examined by RAND, some modifications of the VHSIC Model can be performed to suit the purposes of this report. Following is a brief analysis of the VHSIC model. Later, adjustments to this model will be made based on this analysis and the technology forecasting analogy described in Section 2.3 of this report.

The VHSIC model predicts costs based on perceived relationships of technical parameters to costs. Such a model is typically referred to as a parametric-cost model. A primary advantage of parametric-cost models is that the relationship of technical parameters to the cost is easy to understand. Such models have been used frequently for the evaluation of system costs early in the development of the product life cycle. Some difficulties arise in the number of technical parameters that can be used in parametric cost estimating. Often, model complexity prohibits all possible parameters from being included. Also, accuracy of the cost model depends on the parameters and the data that are available.

The application of the VHSIC model for the estimation of avionics system costs was made relevant because of the data available for the statistical analysis of the model. The

investigators from RAND found that the model developed for avionics computer systems had an  $r^2$  value of 0.88 and a standard error of 0.37. The availability of current data will also play a significant role for the cost estimation of future semiconductor manufacturing. The quantity of data available for the present research does not match up well to the data available to the RAND Corporation. A modified version of a statistically significant model is to be used in this project. By using such a model, it is believed that reliable estimates can be achieved for the future selling price of semiconductor chips.

### **3.1.2 Cost Trends**

Analysis of the business cycle for any industry is likely to produce some type of trend in the cost or demand of the product. In the semiconductor industry, an interesting trend is the exponential decrease in transistor cost. As shown in Figure 2-1, the present cost per transistor on a microprocessor chip has fallen to a couple of millicents. In this analysis, we assume that no major cost trend shifts will occur. For such clearly defined trends, the assumption of continuation for the trend is usually valid. An additional advantage for the use of cost trends is that the extrapolation of consistent, well-defined trends is easy to perform.

A limited view of the manufacturing cost is sometimes obtained, however, when only one parameter is used to predict manufacturing cost. Without other data, the increasing contribution of equipment for the total cost of the factory would not have been known. The inclusion of as many factors as possible is usually desired for more accurate cost estimates. However, model complexity often limits the number of parameters that can be estimated reasonably well. For a more complete picture of industry costs, cost trend data should be used in conjunction with other semiconductor industry parameters such as manufacturing complexity and product performance.

### 3.2 Modification of VHSIC Model

Before any modifications to the VHSIC model are made, the reader should refer back to the original relationship of production cost for VHSIC computer systems to be used in avionics systems. The original Equation 2-2 is given below:

$$T_{100} = 2.20(lbs)^{0.95} (LOI)^{0.07} e^{0.80*(NEW)}$$

Modification of the model given above will be based on those factors that are relevant to the analysis of future manufacturing costs of nanoelectronics technology. The production cost was given as the estimated parameter within the model. For the case of nanoelectronics cost estimation, the value that will be estimated is the initial selling price of the chip. The initial selling price of the chip can often give an indication of the complexity and cost involved with chip manufacture.

The second factor that will be changed is,  $b_0 = 2.20$ , given at the beginning of the equation. This constant was obtained through a series of statistical analyses based on data available to the RAND Corporation. Although the data used by RAND has not been identified, a new numeric factor must be determined based on the data available for this research.

The RAND VHSIC Model was used for the estimation of defense-related computer systems. Often, defense-related products require extreme reliability, extreme service limits, and many quality checks. These costs were included in the RAND Model. All of these stringent factors are usually not of concern for commercial development of chips. The use of the modified model can be used provided that the modified model is in the same form as the generalized model specified by Dean.

A significant consideration in the development of any system used in an aircraft is the weight contribution for that system. The researchers at RAND recognized this important factor and included it in the model that they developed. For the development of future computing processors used in conventional computer systems, weight is not a significant consideration. As such, its inclusion in the model to estimate future manufacturing prices of nanoelectronics technology is not appropriate. Because the exponent for the weight factor was based solely on the contribution of weight to aircraft subsystem costs, its inclusion in the modified cost model is also not appropriate.

Because using the weight factor for the estimation of future semiconductor production costs is not appropriate, it should be replaced by another factor. An appropriate performance parameter that is more indicative of price is the clock speed of the processor. Although this substitution has modified the VHSIC Model, the substitution does not violate the generalized model specified by Dean. Historically, the processing frequency of the chip has been observed to increase with each successive generation of computing devices. By using current data, an estimate for the exponent of the processing frequency factor can be obtained. Sensitivity analysis will be performed on future estimates to predict the factor's relationship to chip price. Current and future processor speed data are available for inclusion in the model. These values are listed below in Table 3-1.

**Table 3-1: High Performance On-Chip Clock Frequency [SIA94]**

| <b>Year</b> | <b>Predicted<br/>Chip Speed (MHz)</b> |
|-------------|---------------------------------------|
| 1998        | 450                                   |
| 2001        | 600                                   |
| 2004        | 800                                   |
| 2007        | 1000                                  |
| 2010        | 1100                                  |

A significant factor in the model developed by RAND was the level of integration factor (LOI). This factor was used to predict system costs based on the number of transistors used on a chip. For the purpose of this research, the inclusion of the LOI factor is desired and its inclusion in the modified model is appropriate. As the exponent for the LOI factor has been determined from multiple analyses of computer system costs, the exponent will remain unchanged in the modified model. A critical assumption made in this project is that the LOI factor will affect commercial system prices in the same way as defense system costs. Sensitivity analysis will be performed on future estimates to predict the exponent's relationship to chip price.

The last factor of the model,  $e^{0.80*(NEW)}$  was used as a parameter to increase the estimated cost depending on the date of technology implementation. The estimated cost would increase by a factor of  $e^{0.80}$ , or 2.23, if the technology within the avionics system was implemented within two years of its first commercial introduction. Otherwise, the estimated cost would only depend on the previous factors included in the model. This binary factor merely provides a constant multiplier for the estimation of the avionics computer system costs examined by RAND. The first factor in the modified model is also a constant. In the modified model, the first factor will compensate for the binary factor given as the last parameter of the original model.

Some critical assumptions have been made for the modification of the VHSIC cost model. First, it is assumed that the replacement of the weight parameter by a clock speed factor does not significantly affect the integrity of the cost estimating relationship. Also, it is assumed that the exponential parameter of the LOI term reflects semiconductor technology advances and not aircraft avionics advances. The constant exponent for the LOI factor is also assumed to remain unaffected by significant technological changes in semiconductor technology. Reusing the exponent for the LOI factor also assumes that commercial system prices will be affected in the same manner as defense-related costs. Finally, it is assumed that

the exclusion of the last term from Equation 2-2 can be reflected in a constant term at the beginning of the modified model. The resulting equation based on the aforementioned assumptions is a special case of general parametric cost estimating model specified by Dean. For a brief discussion of this generic model see Section 2.2.1. The special case of the general model is given:

$$P_{chip} = \alpha (\text{MHz})^{\beta} (\text{LOI})^{0.07} \quad \text{Equation 3-1}$$

where:

- $P_{chip}$  = Initial selling price per chip (\$)
- $\alpha$  = Constant multiplicative factor (no units)
- MHz = Clock speed of the microprocessor chip (MHz)
- $\beta$  = MHz exponent factor (no units)
- LOI = Level of integration (number of transistors/chip)

### 3.3 Extrapolation of Industry Trends

The use of a simple cost trend often does not provide the analyst with a complete view of industry costs. However, a trend used in conjunction with other parameters can be used to give a more complete cost picture. In this section, the cost trend previously identified in Figure 2-1 will be extrapolated to predict future production costs on a per transistor basis. In addition to this prediction, an extrapolation of Figure 1-1 will predict the number of transistor components per microprocessor chip. The product of these two numbers (transistor cost and number of transistors per chip) will predict the future costs of microprocessor chip manufacture.

Extrapolation of the cost per transistor trend was performed using a logarithmic-linear least-squares regression of the data included in Table 2-2. The logarithm (base 10) of the cost data was taken and regressed as a dependent variable. A modified value for the year of each cost data point was included as the independent time variable for the regression. The year was adjusted by using 1977 A.D. as base year 0, 1980 A.D. as year 3, and so on. The result of

this regression provides the cost per transistor for any modified year value. This expression is given by Equation 3-2. The predicted future costs per transistor ( $C_{trans}$ ) are given in Table 3-2, below. Regression output for this data is included in Appendix B of this report.

$$\log_{10}(C_{trans}) = 2.0304 - 0.1067 * (\text{mod year}) \quad \text{Equation 3-2}$$

The expression above is remarkably similar to a regression equation based on the Chow Price Index. This index consists of three measures of computing power (multiplication time, memory size, and access time) [Bro75]. The price index used 1960 as a base year and computes the 1960 rental value computing power in thousands of dollars [Bro75]. All estimates made by the Chow Price Index assume that the rate of price change over time represents the rate of technological progress [Bro75]. The regression of the Chow Price Index on time for the years 1956 to 1965 is shown below:

$$\log_{10} \text{ price} = 1.54 - 0.237 * \text{time} \quad r^2 = 0.99 \quad \text{Equation 3-3}$$

A logarithmic least-squares regression was also performed for the number of transistor components on a computing chip. The data for this regression was taken from historical transistor content of Intel® microprocessor chips used for desktop computing applications. The observed data are included in Appendix B of this report. The logarithm (base 10) of the number of transistors was taken and regressed as a dependent variable. A modified value for the year of each cost data point was included as the independent variable for the regression. The result of this regression is Equation 3-4 and it provides the number of transistors per chip ( $N_{trans}$ ) for any modified year value. A summary of the predictions made by this equation is summarized in Table 3-2. Later, the projections for  $N_{trans}$  will be used in the Modified VHSIC Model. Regression output for this data is included in Appendix B of this report.

$$\log_{10}(N_{trans}) = 4.3407 + 0.1332 * (\text{mod year}) \quad \text{Equation 3-4}$$

From the regression analyses performed, a prediction for the future chip cost ( $C_{chip}$ ) has been calculated. This calculation was performed by taking the product of the predicted

cost per transistor ( $C_{trans}$ ) and predicted number of transistors per chip ( $N_{trans}$ ). The result of this calculation is shown below in Table 3-2. The costs shown predict an increase in chip cost mainly because the pace of increase for transistor content is greater than the ability of manufacturers to reduce costs.

By taking the product of Equations 3-2 and 3-4 a generalized form of the predicted chip cost ( $C_{chip}$ ) produced Equation 3-5:

$$\log_{10}(C_{chip}) = 1.3711 + 0.0265 * (\text{mod year}) \quad \text{Equation 3-5}$$

The cost equation given above provides the chip cost in dollars for the production year of the chip. The multiplicative logarithmic relationship was used to generate Equation 3-5. This relationship states that the product of two terms in logarithmic form is the addition of their respective expressions. In this case, the simple addition yields the cost per chip in millicents. The cost in dollars was obtained by subtracting five orders of magnitude from the resulting expression.

**Table 3-2: Prediction of Future Chip Cost Based on Trend Extrapolation**

| <b>Year</b> | <b>Cost/Transistor<br/>(millicents)</b> | <b>Transistors<br/>per Chip (<math>10^6</math>)</b> | <b>Chip Manufacturing<br/>Cost (\$)</b> |
|-------------|---|---|---|
| <b>1995</b> | 1.286                                   | 5.8991  | 75.86                                   |
| <b>1998</b> | 0.615                                   | 13.7076   | 84.30                                   |
| <b>2001</b> | 0.294                                   | 34.3897   | 101.11                                  |
| <b>2004</b> | 0.141                                   | 86.2771   | 121.65                                  |
| <b>2007</b> | 0.067                                   | 216.4526  | 145.02                                  |
| <b>2010</b> | 0.032                                   | 543.0379  | 173.77                                  |

### **3.4 Statistical Analysis of Time-Series Data**

The relevance of any analysis depends to a large extent on the statistical relevance of the results. The data analyzed in this chapter are the historical transistor cost and observed transistor content for semiconductor computer chips. The accuracy of the estimates for the future manufacturing cost of chips depends on the data and/or the model used to generate the estimate. In this section, two methods for statistically analyzing the data are described. In the first part, output from the regression statistics will be used to describe the linearity and the confidence of the regressed time-series data. The second method of statistical analysis involves testing the data for autocorrelation. Both analyses show that the data and results of the model (i.e., the estimates) are statistically significant with little or no autocorrelation.

#### **3.4.1 Regression of the Time Series Data**

The time-series data for the transistor cost provided a estimate for the transistor cost as a function of time. This estimate was found to be a logarithmic-linear equation. The  $r^2$  for this analysis was found to be 0.996 with a standard error of 0.04. The close value of  $r^2$  to unity indicates that the independent variable and the dependent variable are functionally related. A more telling statistic may be the confidence interval of the coefficients. This accuracy is determined by a statistical measure known as the t-statistic, also known as the Student's t-distribution and ratio. This statistic is the ratio of the estimated coefficient to the computed standard error. The t-statistics for the intercept and the slope of Equation 3-2 are 68.6 and -32.7, respectively. These values of the t-statistic are relevant for six degrees of freedom and a 95% confidence interval. These statistics can be found in Appendix B of this report.

The regressed time-series data for the transistor content also provided a logarithmic-linear equation for the transistor content as a function of time. The value of  $r^2$  for this analysis was found to be 0.991 with a standard error of 0.09. Again, the independent variable and the dependent variable are functionally related. The t-statistics for the intercept and the slope of

Equation 3-4 are 58.5 and -21.4, respectively. These values of the t-statistic are relevant for six degrees of freedom and a 95% confidence interval. These statistics can be found in Appendix B of this report.

### 3.4.2 Autocorrelation for Time Series Data

According to Frank [Fra71], a critical assumption is made with respect to each data point when regressing time-series data. This assumption is that the sample values of the error terms  $e_i$  for  $i = 1, 2, \dots, n$  are independently distributed [Fra71]. If this assumption is violated, the data are said to be autocorrelated. That is, the error term in one period affects the probability distribution of the error term in other periods [Fra71]. Autocorrelation of the data may exist when variables are missing from a relationship or when a non-linear relationship is specified as a linear relationship. Although autocorrelation may not result in biased estimates of the coefficients, the variance of the estimates may be significantly lower because of this effect. Lower values for the variance of the estimate will lead to a higher value of the t-statistic. This higher value may lead the researcher to conclude that the t-statistics are valid, when in fact they are not [Fra71].

To test for autocorrelation, it is necessary to make an assumption about the form of the autocorrelation. A common form of autocorrelation is the simply called linear first-order autocorrelation. One test for first-order autocorrelation is based on the Durbin-Watson statistic  $d$  (This measure is also known as the von Neumann ratio) [Fra71]. The formula for this statistic is given below:

$$d = \frac{\sum_{i=2}^N (e_i - e_{i-1})^2}{\sum_{i=1}^N e_i^2}$$

where  $e_i$  is the estimated error term [Fra71]. If there is no first-order autocorrelation, the expected value of the Durbin-Watson statistic is 2. A high degree of positive autocorrelation

would shift the expected value of  $d$  close to zero. A high degree of negative autocorrelation would shift the expected value of  $d$  close to four [Fra71].

The Durbin-Watson test was performed for the data used to generate Equations 3-2 and 3-4. The data for the test and the numbers used to calculate the Durbin-Watson statistic are given in Appendix B of this report. In the test for the transistor cost data, the Durbin-Watson statistic was found to be 2.38. This value would indicate there is no autocorrelation or a slight negative autocorrelation may exist for the transistor cost data. For the transistor content data, the Durbin-Watson statistic was found have a value of 1.15. This value would tend to indicate that the data may be weakly autocorrelated [Fout96]. This weak correlation, however, would not invalidate the results obtained from the regression analysis [Fout96].

## **CHAPTER 4**

### **RESULTS OF MODIFIED VHSIC MODEL**

In this chapter, initial offering prices of future chips are estimated based on the modifications made to the VHSIC cost model. The initial offering price of semiconductor chips is the price that manufacturers charge for their product after initial development and manufacture. After commercial introduction, the price of such chips is observed to steadily decrease. This decrease in price is a result of learning in the manufacturing process and the introduction of computer chips with greater performance. Although the initial chip offering price is not a direct indicator of the manufacturing cost, this price can be used to identify trends in the manufacturing cost of the chips.

The future estimates for initial chip offering price are obtained by first determining the unknown parameter values ( $\alpha$  and  $\beta$ ) of the modified model. Values for  $\alpha$  and  $\beta$  are obtained by performing a non-linear least square's regression analysis for the Modified VHSIC Model. The performance parameters and initial chip offering price of past and current semiconductor chips are used in the regression analysis. The data obtained for such analysis consists of the initial chip offering price of Intel® chips from the 386, 486, Pentium® and Pentium® Pro generations. Next, future performance parameters of processing frequency and number of transistors on a chip are substituted into the model. Future performance parameters for the number of transistors on a chip are the same projections as those performed in the trend extrapolation section of this report. The substitution of future performance parameters is performed so that a future estimate of initial chip offering price can be obtained. The impact of technology breakthroughs will be considered in terms of their effect on future chip prices.

## 4.1 Parameter Estimation

Considering the development of the modified cost model in Chapter 3, there exist several unknown parameters ( $\alpha$  and  $\beta$ ) within the model. Before any projections of initial chip offering price can be made  $\alpha$  and  $\beta$  must be estimated. The reader may wish to refer back to the modified version of the VHSIC Cost Model developed in Chapter 3 to recall how  $\alpha$  and  $\beta$  are used in the Modified VHSIC Model. The model that has been developed is a special case of the generalized parametric cost model developed by Dean. The original Equation 3-1 is given below:

$$P_{chip} = \alpha (\text{MHz})^{\beta} (\text{LOI})^{0.07}$$

Note that  $\alpha$  is a constant multiplicative factor with no units and  $\beta$  is the exponential factor of the MHz term, also with no units. First, a rough estimate of these parameters is to be obtained by taking the performance parameters and initial price for current and past chips then solving for the unknowns  $\alpha$  and  $\beta$ . Next, future chip price estimates will be obtained by substituting future performance data for semiconductor chips while not changing the estimated parameters. Finally, sensitivity analysis will be performed to determine the dependence of the chip price on the technical parameters included in the model.

The first step is to determine the unknown parameters  $\alpha$  and  $\beta$ . This is performed by using past and current semiconductor chip prices. The historical processing frequency of the microprocessor and the historical number of transistors on the chip are included with the price of the particular chip generation. The prices quoted are the given price of each semiconductor chip for a lot size of 1000. The data are summarized in Table 4-1 below [Int96a,b,c]:

**Table 4-1: Historic Performance and Initial Prices of Intel® Semiconductor Chips**

| <b>Date of Introduction</b> | <b>Intel® Chip</b> | <b>MHz</b> | <b>Number of Transistors</b> | <b>Initial Chip Selling Price (\$)</b> |
|-----------------------------|--------------------|------------|------------------------------|--|
| Oct 1985                    | 386™ DX            | 16         | 275,000                      | 299                                    |
| Sept 1991                   | 486™ SX            | 25         | 1,200,000                    | 349                                    |
| June 1991                   | 486™ DX            | 50         | 1,200,000                    | 665                                    |
| Aug 1992                    | 486™ DX2           | 66         | 1,200,000                    | 682                                    |
| Mar 1993                    | Pentium®           | 66         | 3,100,000                    | 965                                    |
| Mar 1994                    | Pentium®           | 100        | 3,200,000                    | 995                                    |
| Mar 1995                    | Pentium®           | 120        | 3,200,000                    | 935                                    |
| Nov 1995                    | Pentium® Pro       | 166        | 5,500,000                    | 1682                                   |
| Nov 1995                    | Pentium® Pro       | 200        | 5,500,000                    | 1989                                   |

These data are used to obtain the values of the unknown parameters  $\alpha$  and  $\beta$ . A linear least square's regression is performed on the data shown in Table 4-1. The reader may note that Equation 3-1 is not linear. However, by taking the natural log of both sides a linear equation is formed. The following expression is regressed with the left hand side as the dependent variable and the  $\ln(\text{MHz})$  term as the independent variable:

$$\ln\left(\frac{P_{chip}}{LOI^{0.07}}\right) = \ln a + b \ln(\text{MHz})$$

The regression analysis of this data provides estimates for the slope ( $\beta$ ) and intercept ( $\ln \alpha$ ) of the equation stated above. The regression provides the value of  $\beta$  to be 0.660 with a standard error of 0.065. The intercept ( $\ln \alpha$ ) of the regressed linear equation was found to be 2.884 with a standard error of 0.28. To find the value of  $\alpha$ , the inverse natural log was taken of the estimated intercept. This calculation yielded a value of 17.891 for  $\alpha$ . The regression provided an  $r^2$  value of 0.94. The close value of  $r^2$  to unity indicates that the independent variable and the dependent variable are functionally related. The values for the t-statistics of

the estimates of the intercept and slope are 10.30 and 10.15, respectively. These values for the t-statistic are relevant for nine degrees of freedom and a 95% confidence interval. Calculation of the Durbin-Watson Statistic provided a value of 1.83, a number very close to the optimal 2. From this calculation, it can be concluded that the data are not autocorrelated in the first order. These statistics can be found in Appendix C of this report. In Chapter 5, a sensitivity analysis of the projected chip cost is performed for variations in the estimated parameters  $\alpha$  and  $\beta$ .

## 4.2 Predicted Chip Prices

Since the unknown parameters have been obtained from the regression of past and current chip cost data, the modified cost model is complete. This cost model will predict future semiconductor chip prices based on the future semiconductor production parameters of processor frequency and number of transistors on a chip. Regression of the data shown in Table 4-1 results in the Modified VHSIC model is expressed as Equation 4-1:

$$P_{chip} = 17.891(MHz)^{0.66}(LOI)^{0.07} \quad \text{Equation 4-1}$$

where:  $P_{chip}$  = Initial selling price per chip (\$)  
 MHz = Clock speed of the microprocessor chip (MHz)  
 LOI = Level of integration (number of transistors/chip)

By using the expression developed above, the data available for future processing frequency and predictions for the number of transistors on a semiconductor chip, future chip prices were calculated. The projections for the future processing frequency of semiconductor chips was taken from the National Technology Roadmap for Semiconductors [SIA94]. The predictions used for the number of transistors per chip were taken from the extrapolation performed in Section 3.3 of this report. Note that the future chip prices are only estimates based on Intel® data. The application of this model for other chip manufacturers is not valid.

Other manufacturers must re-calculate parameters based on historical costs and historical performance parameters for the particular manufacturer.

The following estimates for chip price include most if not all of the primary cost contributors in a manufacturing facility. These contributors include, but are not limited to, the materials used in fabrication, facility overhead, labor, cost of capital equipment and maintenance costs. The future chip prices as predicted by the Modified VHSIC Model are shown below in Table 4-2. Also included in the table are the performance parameters used in the model and the year for which the performance parameters may be realized.

**Table 4-2: Predicted Chip Price from Modified VHSIC Model**

| <b>Year</b> | <b>MHz [SIA94]</b> | <b>Transistors per Chip (<math>10^6</math>)</b> | <b>Initial Chip Selling Price (\$)</b> |
|-------------|--------------------|---|--|
| 1998        | 450                | 13.708  | 3,200                                  |
| 2001        | 600                | 34.390  | 4,100                                  |
| 2004        | 800                | 86.277  | 5,300                                  |
| 2007        | 1000               | 216.453   | 6,500                                  |
| 2010        | 1100               | 543.038   | 7,400                                  |

The extrapolated estimates made in Chapter 3 are for semiconductor chip manufacturing costs. The projections obtained from the Modified VHSIC Model are for the initial selling prices of the same chips. Although the selling price is predicted in the Modified VHSIC Model and not the manufacturing cost, an observed trend in the selling price of the chips can give an indication of the manufacturing cost trend. Despite that the two measured quantities are not identical, the primary interest in the analysis of these two results is the trends that both these measures predict. Also, the manner in which the trends can explain technological developments that lead to lowered production costs is of interest. The impact of technological developments on future cost is described in Section 4.3 of this report.

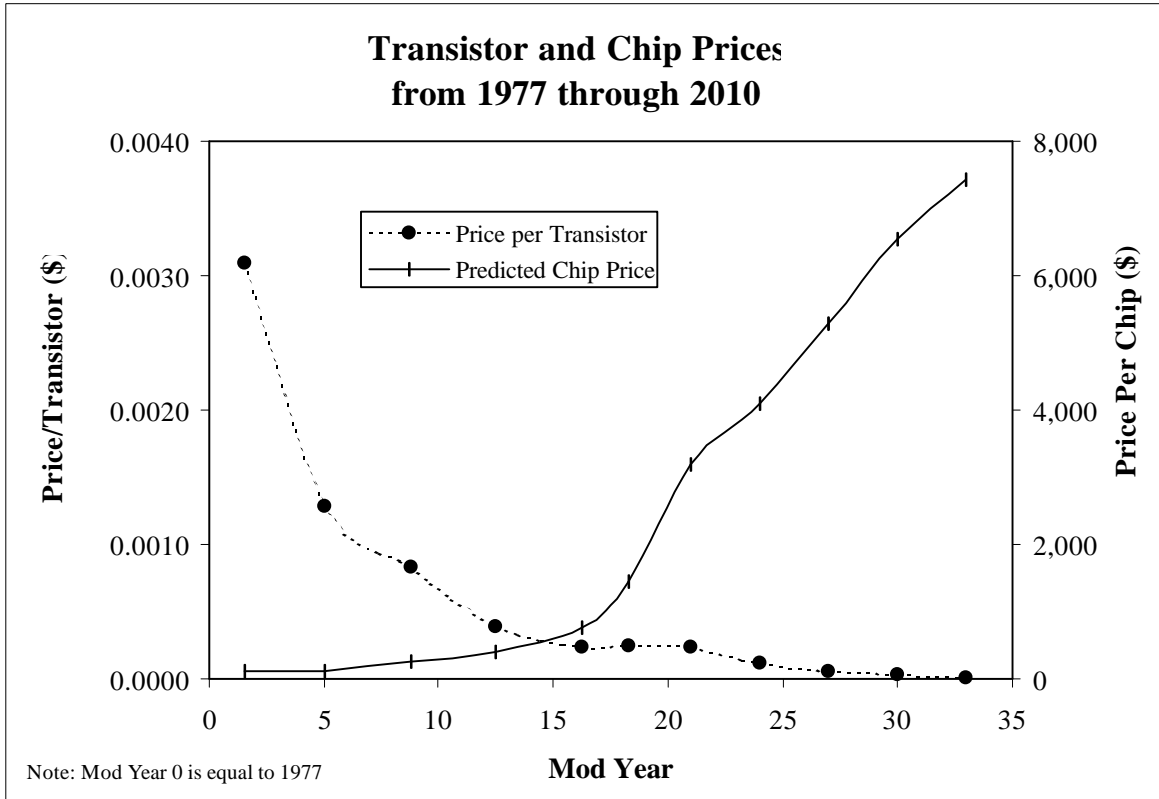
The predicted semiconductor chip prices and the price per transistor are shown below in Figure 4-1. The predicted chip prices are the results of the projections made by the Modified VHSIC Model. The transistor prices were obtained by dividing the chip price by the predicted number of transistors. Chip prices are observed to increase steadily throughout the development of computer chip production. Despite these spiraling prices, the transistor price within each chip generation is shown to decrease over time. This would tend to indicate that chip manufacturers will be able to provide more computing functionality per dollar spent on each subsequent chip generation.

### **4.3 Impact of Technology Breakthroughs**

The results obtained from the Modified VHSIC Price Model generate a similar trend to the results obtained from the extrapolation of industry trends. That trend is that both semiconductor chip manufacturing costs and initial selling prices will steadily continue to rise. These trends were obtained through the assumption of continued industry development and the push towards greater chip performance. However, the impact of technology breakthroughs on system cost may be used to explain how the predicted costs may vary from actual future costs.

Both data sets include assumptions for the future performance and cost of the semiconductor chips. These assumptions are based on observed industry trends and technological breakthroughs that are evolutionary in nature. The trends do not take into account revolutionary changes in the manufacture or design of semiconductor devices. In this respect, the technological breakthroughs that may result from revolutionary changes are ignored. One strong reason for ignoring these changes is that revolutionary shifts in the manufacture or design of semiconductor devices is difficult, if not impossible, to predict. However, just as the invention of solid-state transistor elements replaced the vacuum tube,

other fields of semiconductor design and manufacture may overtake the current methods of production.



**Figure 4-1:** The predicted semiconductor chip prices are shown. The results from the Modified VHSIC Model are shown in conjunction with the predicted price per transistor for each computer chip generation.

A revolutionary change in the manufacturing of semiconductor chips may occur with the devices that will be able to be made with the adoption of x-ray lithography techniques. The future use of x-ray lithography for the exposure of semiconductor wafers appears to be a viable and necessary technology for future computer chip production. Although x-ray lithography may not be revolutionary in nature, the nano-scale devices constructed from this production technique will be. The contributions that these devices will make to the enhanced performance of computing chips may be similar or even greater than the advances achieved from the adoption of solid-state transistors.

The impact of these developments on the cost experienced by manufacturers is explored in a field called technology forecasting which deals with the impact that technology breakthroughs may have on future performance characteristics and system cost. Figure 2-3 (on page 25) demonstrates that technology breakthroughs can have economic and technologic effects on the manufacturing cost of a product. The possible economic benefits of technology breakthroughs are that computing costs may be lowered through enhanced manufacturing techniques. This may not mean that chip prices will decrease, but the overall cost per electronic function will decrease. This effect is clearly shown in Figure 4-1 where chip prices are observed to increase with time, but the price per transistor is observed to sharply decrease. Therefore, the advent of computing technology based on the limitations currently being experienced will not only lead to enhanced performance characteristics, but also to more computing power at a lower price per transistor.

#### **4.4 Summary**

The modifications that were made to the VHSIC Price Model were used to predict future semiconductor chip prices. To perform the estimates of future prices, the unknown parameters were estimated and future performance parameters were placed in the model. The future chip prices predicted by the Modified VHSIC Model were summarized in Table 4-2. The estimated price per chip and the price per transistor were shown in Figure 4-1.

Despite the steady increase in chip price, the price per transistor was observed to decrease for successive semiconductor chip generations. This observation would indicate that manufacturers will continue to provide greater computing function at a lower cost per transistor. Finally, the impact of technology breakthroughs is explored for advancements in manufacturing and design technologies. Technology breakthroughs may contribute to lower chip costs as a result of revolutionary changes in the design and manufacture of chips.

## CHAPTER 5

### ANALYSIS OF PREDICTED PRICES

In this chapter, the results of the Modified VHSIC Price Model are analyzed. The sensitivity of the initial selling price is considered with respect to the technical and estimated parameters of the model. As each parameter of the model is varied by  $\pm 50\%$ , the initial selling price is computed. A profit-time equation is also developed in this chapter based on the equations previously determined. This profit-time equation shows the initial selling price and the manufacturing cost for any given year. The result of this equation is that the profit margin per chip is shown to increase with each new generation of computing. The profit per transistor, however, is shown to decrease for each year. The variation in the selling price for a common Intel® microprocessor will be shown. This graphic will illustrate that the calculated profit margins are reasonable estimates. Finally, the impact of enhanced performance characteristics for computing devices will be considered.

#### 5.1 Sensitivity of Modified VHSIC Model

One-at-a-time sensitivity analysis is performed for the parameters included in the Modified VHSIC model. This sensitivity analysis shows the difference in the chip selling price for  $\pm 50\%$  changes in each parameter of the model. Equation 4-1, for which the parameters are modified, is given below:

$$P_{chip} = 17.891(MHz)^{0.66} (LOI)^{0.07}$$

In the sensitivity analysis, five parameters were modified. These parameters are  $\alpha$ , MHz,  $\beta$ , LOI, and the exponent for LOI (0.07). Two parameters were technical (MHz and LOI), two other parameters ( $\alpha$  and  $\beta$ ) were determined from the linear regression of the model performed in Chapter 3 and the last parameter is the constant exponent of the LOI term (0.07).

Graphs of the price sensitivity for changes in the technical and estimated parameters are given in Appendix D of this report. For each graph, the sensitivity of the selling price is given for years 1998, 2004, and 2010.

Between the two technical parameters (MHz and LOI), the selling price of the chip largely depends on the processing frequency of the chip. The selling price of the chip increased by a factor of 2 within the  $\pm 50\%$  range for the MHz sensitivity. Varying the LOI factor from -50% to +50% resulted in only an 8% increase in the selling price of the chips. Variations in the exponent of the LOI term resulted in significant increases in the initial chip selling price as compared to changes in the MHz and LOI terms. Within the  $\pm 50\%$  range for the LOI exponent, the initial selling price almost increased by a factor of 4. For analyses where more data are available, multiple regression should be performed to more accurately determine the value of the LOI exponent.

Between the estimated model parameters  $\alpha$  (the constant multiplicative factor) and  $\beta$  (the exponent for the MHz parameter), the selling price was more dependent on  $\beta$ . Although the price almost tripled by varying  $\alpha$  through the  $\pm 50\%$  range,  $\beta$  was more affected. For -50% variance in the  $\beta$  parameter, the chip price in 2004 was merely \$584. By increasing the value of  $\beta$  by 50% the chip price rose to over \$48,000.

## **5.2 Profit-Time Equations for Nanoelectronics**

In this section, the cost and price equations previously constructed in Chapters 3 and 4 will be analyzed. This analysis will consist of the derivation of two profit prediction models that are a function of time. These models are different from the Modified VHSIC Price Model in that the previous model is a function of the performance parameters of a semiconductor chip. The first model that will be derived is a profit-time equation on a per-chip basis. The second model will consider profits on a per-transistor basis.

### 5.2.1 Chip Profit Equation

In Chapter 4, an equation was determined based on historical information from Intel® Microprocessor chips. By taking the natural log of both sides of Equation 4-1, the following equation results:

$$\ln price = \ln 17.891 + 0.66 \ln(MHz) + 0.07 \ln(LOI) \quad \text{Equation 5-1}$$

In order to determine the profit, the manufacturing cost of the chips must be considered. The manufacturing cost per chip was determined in Chapter 3 and is expressed in Equation 3-5:

$$\log_{10}(C_{chip}) = 1.3711 + 0.0265 * (\text{mod year})$$

In order to derive an equation for price based only on the year, a relationship must be established for the technical parameters and the year. Such a relationship was found in Chapter 3 for the LOI term. The reader should recall that the (mod year) term of these expressions uses 1977 as year 0, 1980 as year 3, and so on. Equation 3-4 states:

$$\log_{10}(LOI) = 4.3407 + 0.1332 * (\text{mod year})$$

A relationship between the MHz term and the year has yet to be determined. However, the data previously used to determine the estimated parameters of the Modified VHSIC Model can be used to find this relationship. By performing a regression analysis of the MHz data, Equation 5-2 was formed. Regression statistics can be found in Appendix E of this report.

$$\ln MHz = 1.022 + 0.1985(\text{mod year}) \quad \text{Equation 5-2}$$

By combining Equations 3-4, 3-5, 5-1, and 5-2, the following relationship for the profit per chip was obtained:

$$profit_{chip} = e^{(3.5588+0.131(mod\ year)+0.07\ln(10^{(4.3407+0.1332(mod\ year))}))} - 10^{(1.3711+0.0265(mod\ year))}$$

**Equation 5-3**

A graph of Equation 5-3 is shown in Appendix E of this report. The result of this equation is that the profit margin per chip is shown to increase with each new generation of computing. As expected, the graph of this equation is almost identical to the solid line shown in Figure 4-1 of this report. Figure 4-1 was generated using specific data from the production of Intel® Microprocessor Chips. The contribution that this equation gives is that it expresses the profit per chip solely as a function of the year for future computing processors.

### 5.2.2 Transistor Profit Equation

To determine the relationship for profit on a per transistor basis, the equations for selling price and manufacturing cost per transistor must be determined. In Chapter 3 a relationship for the cost per transistor was found. This relationship was stated in Equation 3-2:

$$\log_{10}(C_{trans}) = 2.0304 - 0.1067 * (mod\ year)$$

With the data used previously to determine the parameters  $\alpha$  and  $\beta$ , a relationship was obtained for the selling price per transistor. The regression statistics for this equation can be found in Appendix E of this report.

$$\ln P_{trans} = -5.663 - 0.154(mod\ year) \quad \text{Equation 5-4}$$

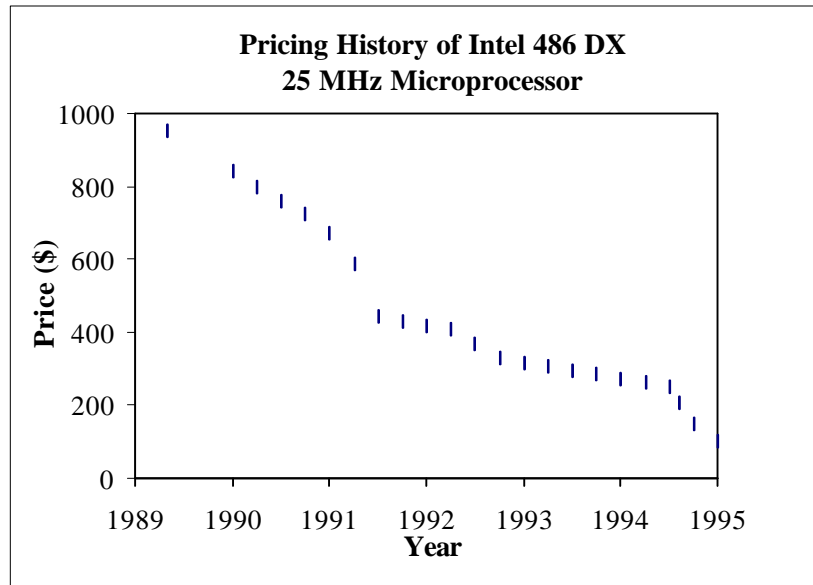
By combining the equations stated above, the following relationship for the profit per transistor is obtained:

$$profit_{trans} = e^{(-5.663-0.154(mod\ year))} - 10^{(2.0304-1.067(mod\ year))} \quad \text{Equation 5-5}$$

The result of this equation is that the profit margin per transistor is shown to decrease with each new generation of computing. As expected, the graph of this equation is also nearly identical to the dashed line shown in Figure 4-1 of this report. Figure 4-1 was generated using specific data from the production of Intel® Microprocessor Chips. The contribution that this equation gives is that it expresses the profit per transistor solely as a function of the year for future computing processors.

### **5.3 Price-Drop Analysis**

An analysis of the selling price for the Intel® chips reveals that there is a significant price-drop throughout the life of the chip. This price-drop may indicate that the profit margins in the semiconductor industry are quite large. The predictions made by Equation 5-3 indicate that the ratio of initial selling price and manufacturing cost can sometimes be as high as forty (40). This number may seem to be unreasonably high, but the reader should consider the historical pricing performance of Intel® chips. Shown in Figure 5-1 is the historical selling price of a popular Intel® chip introduced in April 1989. The initial selling price of the 486™ DX 25 MHz microprocessor was \$950. New, high-performance chips decreased the demand for the chip after several years. As demand decreased so did the price until the steady-state selling price of \$102 per chip was achieved in Q1 1995.



**Figure 5-1:** The historical selling price for the Intel® 486 DX 25 MHz chip is shown. The selling price is observed to decrease by almost a factor of 10 throughout the useful life of the chip. Note the sharp decrease in price in the middle of 1991 as a result of the introduction of the Intel® 486 DX 50 MHz chip.

## 5.4 Enhanced Computing

At this time, there are developments in the research of computing that may lead to revolutionary advances. These advances may lead to enhanced computing devices that will be able to solve previously unsolved problems. Just as the advent of the solid-state transistor revolutionized computing, a new transistor design or manufacturing process may break the trends extrapolated in this report. These technology breakthroughs may serve to decrease the manufacturing cost or may increase the functionality of the computing chip.

## **CHAPTER 6**

### **CONCLUSIONS AND RECOMMENDATIONS**

This report has attempted to quantify manufacturing costs and initial selling prices for the future production of more advanced transistor electronics. These estimates have been obtained through the use of trend extrapolation, technology forecasting, and a modified parametric price model. The conclusions of this research are as follows:

- Extrapolation of transistor cost trends showed that the price per electronic component will continue to decrease at a rate of 50% every three years. This extrapolation was performed using regression analysis and was statistically verified using correlation coefficients and t-statistics.
- Extrapolation of chip transistor content showed that the level of integration will double every three years. This estimate is a conservative estimate when compared to Moore's Law. This extrapolation was performed using regression analysis and was statistically verified using correlation coefficients and t-statistics.
- Manufacturing costs for semiconductor chips will continue to rise approximately 6% per year. This estimate is based on the continuation of the trends extrapolated for transistor cost and transistor content.
- Modification of a generalized parametric cost model yielded a useful price model for the prediction of future chip selling prices.

- Future chip selling prices will increase at a rate of approximately 8% per year over the next 15 years. This estimate is based on the continuation of established microelectronics industry trends and technology forecasting.
- Profit-time equations were developed that predict industry profits on a per-chip and per-transistor basis. Chip profits were observed to increase at roughly the same rate of selling prices, and transistor profits were observed to decrease at roughly the same rate of selling prices.

## **6.1 Relevance**

Despite the decrease in transistor costs predicted in this report, this work anticipates that the costs associated with facility and equipment development will continue to skyrocket! These spiraling costs will necessitate more cost-effective solutions for the manufacture of future computing devices.

One such proposal is the implementation of x-ray lithography for the fabrication of chips that take advantage of the limitations currently being encountered. Although x-ray lithography will not be feasible for industrial use until several years, or decades in the future, it may offer a means to overcome some of the cost-hurdles for future computer chip manufacture. In order for industry to further justify capital investment decisions, cheaper methodologies for manufacture of computer chips must be implemented, or performance of these chips must continue to increase at a faster rate than costs.

## **6.2 Recommendations for Future Research**

The Modified VHSIC Price Model only considers two technical parameters for the estimation of initial chip selling price. These two parameters are the processing speed of the chip (MHz) and the number of transistors on the chip (LOI). Other factors such as the minimum feature size of the chip and the bus width may be critical metrics in the estimation

of future costs. Future research may consider the inclusion of more technical parameters to generate a more complete model.

Validation of the Modified VHSIC Model has yet to be performed. Validation of this model might be performed by using data available from a chip manufacturer such as Motorola. Motorola may be a good choice for model validation because the primary focus of chip development is similar to Intel® Corporation. In addition, Motorola's manufacturing progress in the number of transistors per chip has been nearly identical to the Intel® Corporation [Hutch96]. An even more accurate model could also be formulated by using data from both companies.

The results produced in this report apply strictly to the silicon-based microprocessor computing industry. Other technologies such as gallium arsenide, memory, and application specific integrated circuits (ASICs) have not been considered herein.

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## **APPENDIX A**

### **TRANSISTOR ELECTRONICS AND ITS FUTURE**

The following section supplements Chapter 1 by reviewing in greater depth some of the more technical aspects associated with the further miniaturization of transistor electronics [Monte95], [Nanos95]. In particular, the origin and fundamental concepts of the transistor are discussed. Applications for the transistor and the advantages of a solid-state design instead of the vacuum tube predecessor are addressed. The development and application of transistor designs for computers and computing are highlighted. A pictorial description of the photolithography fabrication process discussed in Section 1.3.1 is included. A full description of the physical limitations for the miniaturization of current microelectronics designs is also included.

As nanotechnology is being explored within several scientific fields, the research efforts within mechanical, chemical, quantum, and electrical fields are described. Two possible devices to be constructed on the nano-scale are described. A description of a Scanning Tunneling Microscope as a nanoelectronic fabrication technology is discussed. Finally, several predictions are made at the end of this chapter concerning the most likely path of development for nanoelectronics technology.

#### **A.1 The Transistor**

This century's most important invention, the transistor, has led to the invention of resources that are essential to the entire population. It is the maturation of transistor fabrication technology throughout the century that has allowed for more complex resources based on transistor technologies. In this section, the invention of the transistor and its development are discussed. The application of transistor, both in the past, current, and future

are addressed. Finally, the development of the transistor for computing applications is examined.

### **A.1.1 Birth of Electronics**

The control of the flow of electrons within a vacuum tube is what first interested a young electrical engineer named Lee De Forest. He found that if an electrified wire grid was placed across a stream of electrons within the vacuum tube, the electron flow could be interrupted, reduced, or stopped entirely [Rock48]. Also, De Forrest found that this device could be used to “amplify” a weak current of electrons flowing in one end of the vacuum tube to a strong current at the outgoing end of the tube [Rock48]. At the beginning of the century, this device was used to give birth to the technology of electronics. Future developments, however, proved to be equally significant in the growth of the electronics.

### **A.1.2 Vacuum Tube Modifications and Applications**

Although the implementation of the vacuum tube in the electronics field was very significant, the control of electrons within a solid-state device has proven to be advantageous because of the miniaturization of the solid-state design. The invention of the transistor has given birth to radio, television, radar, electron microscopes, electronic calculators, robot manipulation, and most notably the computer. Through this modification, the complication and delicacies of the vacuum tube have been eliminated. The transistor does not need to be warmed up allowing the solid state device to be used immediately [Rock48]. Also the power requirements for the transistor are very small. The utilization of more transistors within an electronic device is also made possible through the ability of the transistor to be made less expensive. The ability of mass production of transistors within a single device produced some of the most significant developments within the field of computing.

### **A.1.3 Development of Solid State Computing**

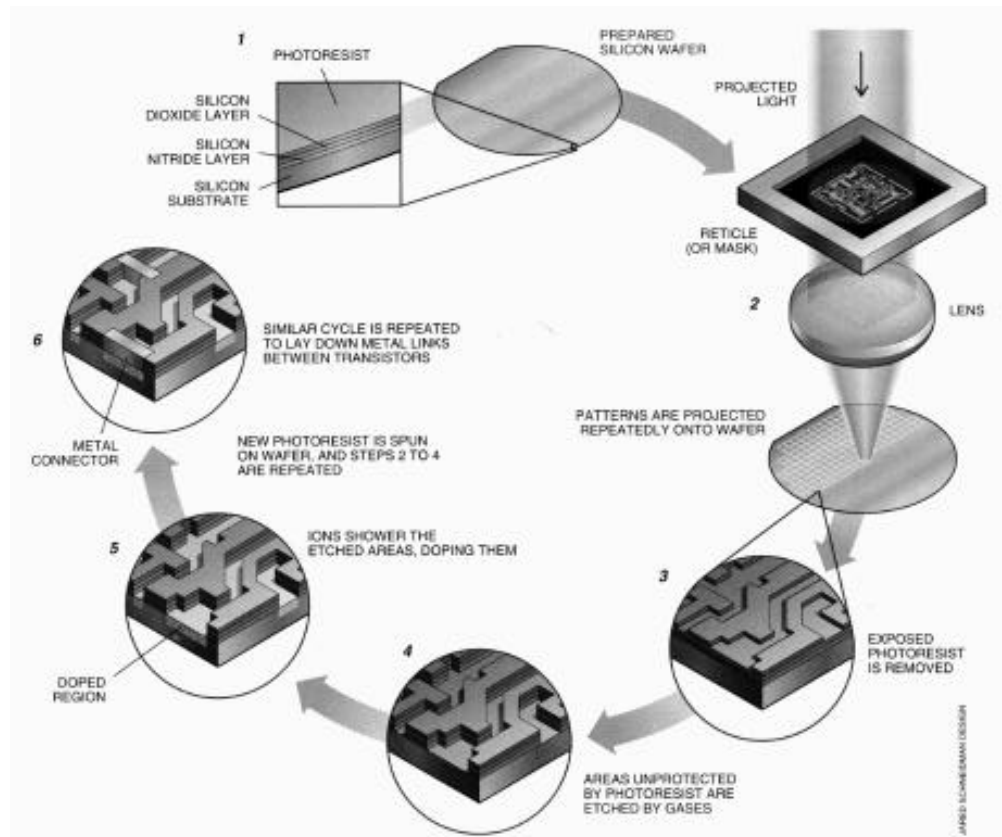
In the early stages of the development of the microelectronics industry, the driving force for advancement was the persistent and rapid decline of cost for a given electronic function. As production within the industry doubled, the integrated circuit costs were observed to decrease 20 to 30 percent [Noyce77]. The main technical barrier for industry development was the production yield. As circuit complexity increased, the probability of circuit defects increased. The most cost effective design resulted from a compromise between the high assembly costs of chips with low levels of integration and the high scrap costs of high levels of integration [Noyce77].

## **A.2 Semiconductor Manufacturing and its Limitations**

As manufacturers attempt to create computer chips with smaller feature sizes, it becomes more difficult for lithography hardware to precisely image the circuit pattern on the chip. A pictorial description of the photolithography process is shown below in Figure A-1. As with any manufacturing process, there exist limitations that affect the performance and the yield of satisfactory products from that process. As the limitations that were previously experienced for photolithography were overcome, several new limitations have emerged. A brief description of these additional obstacles is included in this section of the Appendix.

### **A.2.1 Semiconductor Manufacturing**

Processes for the fabrication of microprocessors have been dominated by photolithography. Photolithography is a printing process similar to “screen printing” for the application of designs on a T-shirt. The lithography printing process was first invented in the 18th Century by a German map inspector [Stix95]. The photolithography fabrication process as described in Section 1.3.1 of this report is shown below in Figure A-1:



**Figure A-1:** The process shown occurs as much as 20 times during chip fabrication. Up to 200 chips can be fabricated on a single wafer. (1) Photoresist is applied. (2) Each cycle has a unique pattern that is projected repeatedly on the wafer. (3) Photoresist is removed. (4) Light exposed areas are etched by gases. (5) Etched areas are showered with dopants creating transistors. (6) Transistors are connected as the next layers add metal and insulation. [Hutch96]

### A.2.2 Physical Limitations

Current microelectronics fabrication processes are able to produce many computer chips with high yields and a high level of circuit integration. The obstacles previously overcome by chip manufacturers are meticulous attention to process control, advancements in clean-room control, and process improvements in photolithography [Noyce77]. As manufacturers attempt to create circuits with more complexity, new technical barriers in the operation of these circuits must be addressed. These new barriers are discussed below.

#### *A.2.2.1 Wavelength of Emitted Light*

Ultra-violet lithography can produce feature sizes only as small as the wavelength of the UV light (usually 350 nanometers). To obtain greater resolution for images on the photoresist, the wavelength of the emitted light must decrease. Semiconductor technology has seen a steady decrease in the wavelength of emitted light used to produce chips. The necessary decreases in the wavelength of emitted light have produced new light sources capable of emitting light at 0.193 microns [Stix95]. Unfortunately, photoresists exposed to the light sources capable of emitting light in the deep-ultraviolet range of the electromagnetic spectrum (less than 0.2 microns) absorb a significant amount of the light [Stix95]. The photoresists absorb so much of the light that it takes more time to transfer the circuit pattern to the chip.

#### *A.2.2.2 Lens Focus on Photoresist*

As the feature sizes of the transistor become smaller, new lens systems must be developed to reduce the image passed through the mask. Also, the ability of this image to be focused clearly on the photoresist is affected. As a result, the aperture of the lens system must be increased. A wide lens aperture increases the resolution but decreases the field of focus. In fact, projected images tend to fade at distances below 1 micron [Stix95]. The inability of the lens to focus to such precise definition on the photoresist is largely a materials problem. The few lens materials that have been identified for the sub-0.2 micron feature size development perform poorly [Stix95]. The fused silica glass for lenses tends to absorb the emitted light and heat up. This increase in temperature of the lens degrades the image by changing the angle at which the lens refracts the light.

#### *A.2.2.3 Tunneling*

Transistor components are separated from each other such that the performance of one component will not affect the performance of the other. For greater transistor densities, the insulation layer between components must decrease. This decrease in insulation layer size

can affect system performance in that electrons can migrate through the insulation layer. The migration of electrons through small energy barriers in electronic circuits is known as tunneling. The result of tunneling is that a transistor that should normally be in an “off state” may in fact be in an “on state” [Bagg92]. An insulation layer more resistant to tunneling may be achieved through the development of materials with a lower dielectric constant.

#### *A.2.2.4 Heat Dissipation*

Current microprocessor chips, such as Intel’s Pentium chip, contain over 3.2 million transistors [SIA94]. While the amount of transistors on a chip has dramatically increased, the size of a computer chip has remained relatively unchanged. The density of components on a chip has a dramatic effect on the heat that must be dissipated from the surface of the chip. If not for cooling fins on the surfaces of today’s microprocessors, many of them would melt themselves down. Conventional transistor designs are eventually going to be limited in density based on the ability of chip manufacturers to siphon heat from the surface of the chip [Keyes77].

### **A.3 Nanotechnology**

Although this project focuses on the development of nanotechnology from the perspective of electronics and computer electronics development, nanotechnology is being explored in many fields of research. The following section will describe some of the implementation technologies for the development of nanotechnology. The devices that are being proposed for the construction of nanoelectronic circuits are also described. Finally, the use of x-ray lithography and Scanning Tunneling Microscope (STM) for the fabrication of nanoelectronic circuits will be highlighted.

#### **A.3.1 Implementation Technologies**

Nanotechnology has proven to be a very diverse field that may obtain applications within the fields of biology, materials, transportation, communications, and computing. In no

other field, however, has there been more interest in the possibilities of nanotechnology than computing. As the minimum feature size within computing devices has decreased, limitations in fabrication techniques have been and continue to be an obstacle for future miniaturization. The technology that will contribute to the development of nano-computing could arise from quantum, mechanical, chemical, and electronic technologies. In this section, each of these technologies is discussed and the advantages and disadvantages for each of these technologies is considered.

#### *A.3.1.1 Quantum*

For the development of computing devices based on the quantum nature of matter, the natural mechanism of interference of quantum waves associated with nano-scale components could be used to perform parallel computations [Bagg92]. Proposed quantum computers would represent a bit of information as a quantum state of some component of the computer, such as the spin orientation of an atom [Bagg92]. The development of computers based on these types of quantum effects is still in a research phase. It is perceived, however, because quantum devices would be very sensitive to small physical distortions and stray photons, such devices must be carefully isolated and operated at temperatures close to absolute zero [Bagg92]. Because quantum nanocomputers could perform massively parallel computations, a primary interest in the application of these devices is to solve certain classes of previously unsolved problems.

#### *A.3.1.2 Mechanical*

The notion of a mechanical device for performing calculations has existed for several centuries. Probably the first mechanical computer was the abacus, created by the Chinese. On the nano-scale, the implementation of a computing device was partly inspired by the macroscopic mechanical computers created by Charles Babbage in the 1830s and 1840s. Mechanical nanocomputers would calculate using moving molecular-scale rods and rotating molecular-scale wheels, that spin on shafts and bearings [Drex92].

A leading authority of nanotechnology, K. Eric Drexler, envisions that such small-scale devices would be assembled by mechanical positioning of atoms or molecules one unit at a time. This process is known as “mechanosynthesis”. A viable technique for mechanosynthesis is the implementation of an STM (Scanning Tunneling Microscope). A major disadvantage for the use of STM is that, as a manipulator of atoms one at a time, fabrication of the even smallest device would prove to be time consuming. The development of STM arrays, however, can provide the ability to build nano-scale components in parallel [Drex92].

#### *A.3.1.3 Chemical*

The ability of making and breaking bonds which process information is the idea behind a chemical computer. Logic states and information are contained within the resulting chemical structure. Proponents of “biochemically-based” computers point to an existence proof that performs computations through the multi-cellular nervous system of humans and other animals. Despite the difficulties in reproducing a living computational system, there has been some evidence of the use of chemical systems for the solution to specific problems. A scientist named Adleman used fragments of DNA to compute the solution to a complex graph theory problem [Adle94]. The greatest potential for application of chemical computations lies in the realm of combinatorial type problems.

#### *A.3.1.4 Electrical*

The development of electronic-based computing elements spans over 50 years. The large infrastructure that has already miniaturized circuit elements is the most poised and mature industry to develop and fabricate nano-scale components. As electronics-based computers have become more popular, the demands for faster and more powerful devices continues to increase. As Figure 1-1 has shown, the rate of increase in the density of transistors on a single chip is exponential. Over that time span, and up to the present, the

fundamental operating principles of the transistor have remained the same. To maintain the current rate of development and miniaturization, changes must occur in the construct of electronic transistors. These changes will occur in manufacturing of such devices and the structure of the components for such devices.

### **A.3.2 Nanoelectronic Devices**

The development of an electronic nanocomputer will represent the storage and retrieval of information by the movement of electrons. To control electrons on a nano-scale, many of the nanodevices proposed to replace the transistor will take advantage of the quantum effects currently experienced by shrinking of bulk-effect transistors. Two such devices are quantum dots and Resonant Tunneling Diodes. These devices are described in the two sections below.

#### *A.3.2.1 Quantum Dots*

Quantum dots control the tunneling of a small number of electrons through the influence of a generated electric field. The dots have shown the capability to perform both switching and amplification characteristics. For an electron, the quantum dot serves as a potential well from which the electron is not free to move in any dimension. For computing purposes, the presence or absence of an electron within the quantum dot can function as a switch (1 or 0) [Horg94]. Some quantum dots have been fabricated to be as small as 30 nanometers.

A group at the University of Notre Dame has proposed and experimented with a method for performing logic functions with arrays of quantum dot cells. These arrays function as a result of the repulsion of like charges with neighboring arrays [Lent93]. This repulsion causes adjacent quantum dot cells to line up in the same state, since the lowest energy for the entire system is achieved in this state. By setting up the appropriate patterns for these arrays of quantum dot cells, Boolean logic functions can be implemented [Lent93].

Since these quantum dot arrays communicate via their electric field and not current flow, their use as a form of wireless computation is possible.

#### *A.3.2.2 Resonant Tunneling Diodes (RTDs)*

The technology for the creation of resonant tunneling diodes has been used to fabricate and experiment with this possible implementation of nanoelectronic devices. An RTD is composed of two insulating barriers in a semiconductor. This creates an island as small as 10 nanometers in width between the two semiconductors [Bate88]. Whenever electrons are confined within two such closely spaced barriers, the quantum wave properties of the electrons restrict the particle energies to discrete energy levels. This quantization of the electrons within energy band is the principle upon which the RTD operates [Bate88]. The probability for the electrons to tunnel from outside the barriers is dependent on the average energy of the incoming electrons. When the average energy of such electrons is equal to the quantized energy levels of the device, significant tunneling can occur.

By incorporating an RTD into the emitter of a BJT (bipolar junction transistor) a “resonant tunneling transistor” has been fabricated. This compound device is controlled by the RTD which serves as a filter to allow current to flow into the emitter of the BJT at certain base-emitter voltages. These gate voltages correspond to specific internal energy levels of the RTD [Bate88]. As the base-emitter voltage increases, the current can go from an “off” state where the base-emitter voltage does not correspond to an internal energy level of the RTD to an “on” state where the internal energy level of the RTD matches the base-emitter voltage. Used in this manner, the resonant tunneling transistor can be used as a two-state device capable of computations.

### **A.3.3 Nanoelectronic Fabrication Technologies**

Regardless of the conceptual inventions for the implementation of nano-scale electronic devices, their eventual application and adoption into mainstream electronics will

depend on the fabrication technologies that can build such devices. As conventional lithography techniques are becoming more expensive and complicated, new manufacturing technologies have emerged. Also, the modification of existing lithography technologies has been explored for the fabrication of electronic devices on the nano-scale. Two such technologies are discussed in this section.

#### *A.3.3.1 X-Ray Lithography*

As semiconductor chip manufacturers move towards smaller wavelengths of the light spectrum, the available visible wavelengths that can be used for the fabrication of microelectronics circuits decreases. This limit has caused several researchers in the microelectronics field to research non-visible forms of electromagnetic waves for lithography. A brief description of x-ray lithography is included in Section 1.3.3.1 of this report.

#### *A.3.3.2 STM*

A radically different method for the manipulation of atoms or molecules has been proposed. This method involves mechanosynthesis through the use of an STM (Scanning Tunneling Microscope). An early example for the use of an STM was the work performed by Eigler and Schweitzer at IBM. This pair used the microscope to position atoms to spell “IBM” [Eig90]. The STM uses a sharp tungsten tip through which a few tenths of a volt can pass to a conducting sample. Tunneling of atoms from the surface of the conducting sample to the tip can occur when the tip is close (0.5 nanometers) to touching the sample [Eig90].

For the fabrication of devices based on STM technology, the tip voltage can be adjusted such that a covalently-bonded atom can move from the surface to the tip. Removal of the atom from the tip can be performed by adjusting the tip voltage. A major problem for the permanent placement of the atoms on the surface of the sample is that at high temperatures the atom may tend to migrate [Eig90], thus altering the structure originally created. The “IBM” that Eigler and Schweitzer created was performed at temperatures near

zero degrees Kelvin. For manipulation of a greater quantity of atoms, an STM array has also been proposed. This device is similar to an STM, however, an array of tips has been constructed to manipulate the atoms on an atomic scale. At this point, STM technology is still very developmental and fabrication has proven to be a very time consuming process.

#### **A.4 Nanoelectronic Developments**

As can be seen from this section, the development and progress within the transistor-based electronics industry has been nothing short of phenomenal. The developments that have occurred over the past five decades have provided the world with more computational power than ever considered possible. The steady development of the electronics industry is also expected to continue well into the next century. To maintain the same rate of development, new technologies must be implemented to overcome the physical and economic barriers that are currently being experienced. These new technologies are expected to take advantage of the current limitations of microelectronics device fabrication. Nanoelectronics development will not only take advantage of these limitations, but may provide enhanced operational characteristics.

Modifications to the lithography process may produce nano-scale devices with high product yield and high reliability. One particular lithography method of interest is x-ray lithography. This technology is limited by the depth of field for the x-rays and the cost of the synchrotrons necessary to generate the x-rays. Also necessary for the successful large-scale development of x-rays are masks and photoresists used in the processing of the chips. The ability of these devices to produce very small minimum feature sizes is the greatest advantage of this technology. The possibility for x-ray devices to produce nano-scale devices for large scale production of electronic devices has been the primary consideration of this research.

**APPENDIX B**  
**REGRESSION STATISTICS FOR**  
**TIME-SERIES DATA**

On the following two pages are the statistics for the regression performed in Chapter 3 of this report. The current cost data and predicted future costs are shown in Table B-1. The current and future chip transistor content is shown in Table B-2.

Statistical analysis to test the data for auto-correlation was performed using the Durbin-Watson, linear first-order autocorrelation. The data used for the analysis as well as the output from the calculation is shown on the last two pages of Appendix B.

**Table B-1: Least-Squares Linear Regression of Transistor Cost Data**

| Year | Mod Year | Observed Cost (millicents) | Observed Cost Log(10) Cost | Regression Results Log(10) Cost | Predicted Cost (millicents) |
|------|----------|----------------------------|----------------------------|---------------------------------|-----------------------------|
| 1977 | 0        | 112                        | 2.049                      | 2.030                           | 107.25                      |
| 1980 | 3        | 47                         | 1.672                      | 1.710                           | 51.31                       |
| 1983 | 6        | 24                         | 1.380                      | 1.390                           | 24.55                       |
| 1986 | 9        | 13.5                       | 1.130                      | 1.070                           | 11.74                       |
| 1989 | 12       | 5.2                        | 0.716                      | 0.750                           | 5.618                       |
| 1992 | 15       | 2.7                        | 0.431                      | 0.429                           | 2.687                       |
| 1995 | 18       | -                          | -                          | 0.109                           | 1.286                       |
| 1998 | 21       | -                          | -                          | -0.211                          | 0.615                       |
| 2001 | 24       | -                          | -                          | -0.531                          | 0.294                       |
| 2004 | 27       | -                          | -                          | -0.852                          | 0.141                       |
| 2007 | 30       | -                          | -                          | -1.172                          | 0.067                       |
| 2010 | 33       | -                          | -                          | -1.492                          | 0.032                       |

**SUMMARY OUTPUT**

| <i>Regression Statistics</i> |         |
|------------------------------|---------|
| Multiple R                   | 0.99814 |
| R Square                     | 0.99628 |
| Adj. R Square                | 0.99535 |
| Standard Error               | 0.04092 |
| Observations                 | 6       |

**ANOVA**

|            | <i>df</i> | <i>SS</i> | <i>MS</i> | <i>F</i> | <i>Significance F</i> |
|------------|-----------|-----------|-----------|----------|-----------------------|
| Regression | 1         | 1.79438   | 1.79438   | 1071.490 | 5.194E-06             |
| Residual   | 4         | 0.00670   | 0.00167   |          |                       |
| Total      | 5         | 1.80108   |           |          |                       |

|              | <i>Coefficients</i> | <i>Standard Error</i> | <i>t Stat</i> | <i>P-value</i> |
|--------------|---------------------|-----------------------|---------------|----------------|
| Intercept    | 2.0304              | 0.0296                | 68.5539       | 271.27E-9      |
| X Variable 1 | -0.1067             | 0.0033                | -32.7336      | 5.19E-6        |

|  | <i>Lower 95%</i> | <i>Upper 95%</i> |
|--|------------------|------------------|
|  | 1.9482           | 2.1126           |
|  | -0.1158          | -0.0977          |

**Table B-2: Least-Squares Linear Regression of Transistor Content Data**

| Year | Mod Year | Observed Transistors (N) | Observed N Log(10) N | Regression Results Log(10) N | Predicted N |
|------|----------|--------------------------|----------------------|------------------------------|-------------|
| 1978 | 1.5      | 29,000                   | 4.4624               | 4.5404                       | 34.71E+3    |
| 1982 | 5        | 134,000                  | 5.1271               | 5.0065                       | 101.50E+3   |
| 1985 | 8.8      | 275,000                  | 5.4393               | 5.5125                       | 325.43E+3   |
| 1989 | 12.5     | 1,200,000                | 6.0792               | 6.0051                       | 1.01E+6     |
| 1993 | 16.25    | 3,100,000                | 6.4914               | 6.5045                       | 3.19E+6     |
| 1995 | 18.25    | 5,500,000                | 6.7404               | 6.7708                       | 5.90E+6     |
| 1998 | 21       | -                        | -                    | 7.1370                       | 13.71E+6    |
| 2001 | 24       | -                        | -                    | 7.5364                       | 34.39E+6    |
| 2004 | 27       | -                        | -                    | 7.9359                       | 86.28E+6    |
| 2007 | 30       | -                        | -                    | 8.3354                       | 216.45E+6   |
| 2010 | 33       | -                        | -                    | 8.7348                       | 543.04E+6   |

SUMMARY OUTPUT

| <i>Regression Statistics</i> |         |
|------------------------------|---------|
| Multiple R                   | 0.99568 |
| R Square                     | 0.99138 |
| Adj. R Square                | 0.98922 |
| Standard Error               | 0.09023 |
| Observations                 | 6       |

ANOVA

|            | <i>df</i> | <i>SS</i> | <i>MS</i> | <i>F</i> | <i>Significance F</i> |
|------------|-----------|-----------|-----------|----------|-----------------------|
| Regression | 1         | 3.74438   | 3.74438   | 459.877  | 28.0E-6               |
| Residual   | 4         | 0.03257   | 0.00814   |          |                       |
| Total      | 5         | 3.77695   |           |          |                       |

|              | <i>Coefficients</i> | <i>Standard Error</i> | <i>t Stat</i> | <i>P-value</i> |
|--------------|---------------------|-----------------------|---------------|----------------|
| Intercept    | 4.3407              | 0.0743                | 58.4568       | 512.82E-9      |
| X Variable 1 | 0.1332              | 0.0062                | 21.4447       | 27.96E-6       |

|  | <i>Lower 95%</i> | <i>Upper 95%</i> |
|--|------------------|------------------|
|  | 4.1345           | 4.5469           |
|  | 0.1159           | 0.1504           |

## Durbin-Watson Calculation of Linear First Order Auto-Correlation

### Part 1: Auto-Correlation Analysis of Transistor Cost Data

| $Y_i$         | $Y_{ip}$       | $e_i = Y_i - Y_{ip}$ |
|---------------|----------------|----------------------|
| Observed Cost | Predicted Cost | Estimated Error      |
| 112           | 107.25         | 4.749                |
| 47            | 51.31          | -4.309               |
| 24            | 24.55          | -0.546               |
| 13.5          | 11.74          | 1.757                |
| 5.2           | 5.62           | -0.418               |
| 2.7           | 2.69           | 0.013                |

| $e_i - e_{i-1}$                  | $(e_i - e_{i-1})^2$   | $(e_i)^2$                    |
|----------------------------------|-----------------------|------------------------------|
| Difference of<br>Adjacent Errors | Difference<br>Squared | Square of<br>Estimated Error |
| -                                | -                     | 22.551                       |
| -9.057                           | 82.036                | 18.564                       |
| 3.763                            | 14.158                | 0.298                        |
| 2.303                            | 5.305                 | 3.088                        |
| -2.175                           | 4.731                 | 0.174                        |
| 0.430                            | 0.185                 | 0.000                        |

| Sum of Diff.<br>Squared | Sum of Square<br>of Est Error |
|-------------------------|-------------------------------|
| 106.4143                | 44.6756                       |

| Durbin-Watson<br>Statistic |
|----------------------------|
| 2.38                       |

## Durbin-Watson Calculation of Linear First Order Auto-Correlation

### Part 2: Auto-Correlation Analysis of Transistor Content Data

| $Y_i$                   | $Y_{ip}$                   | $e_i = Y_i - Y_{ip}$ |
|-------------------------|----------------------------|----------------------|
| Actual # of Transistors | Predicted # of Transistors | Estimated Error      |
| 29,000                  | 34,707                     | 5707                 |
| 134,000                 | 101,501                    | -32499               |
| 275,000                 | 325,432                    | 50432                |
| 1,200,000               | 1,011,898                  | -188102              |
| 3,100,000               | 3,194,999                  | 94999                |
| 5,500,000               | 5,899,054                  | 399054               |

| $e_i - e_{i-1}$               | $(e_i - e_{i-1})^2$ | $(e_i)^2$                 |
|-------------------------------|---------------------|---------------------------|
| Difference of Adjacent Errors | Difference Squared  | Square of Estimated Error |
| -                             | -                   | 32.58E+6                  |
| -38207                        | 1.46E+9             | 1.06E+9                   |
| 82931                         | 6.88E+9             | 2.54E+9                   |
| -238534                       | 56.90E+9            | 35.38E+9                  |
| 283102                        | 80.15E+9            | 9.02E+9                   |
| 304054                        | 92.45E+9            | 159.24E+9                 |

| Sum of Diff. Squared | Sum of Square of Est Error |
|----------------------|----------------------------|
| 237.83E+9            | 207.28E+9                  |

| Durbin-Watson Statistic |
|-------------------------|
| 1.15                    |

**APPENDIX C**

**REGRESSION STATISTICS FOR  
PARAMETER ESTIMATION**

The following data are the regression statistics of historical prices for Intel® Semiconductor chips. This data was used for the estimation of the parameters of the Modified VHSIC Model. Calculation of the Durbin-Watson Statistic is also performed in this Appendix. This calculation verifies that the data are not autocorrelated in the first-order.

**Table C-1: Historical Intel® Chips: Performance and Price**

| <b>Date</b> | <b>Intel Chip</b> | <b>MHz</b> | <b># of Trans</b> | <b>Price (\$)</b> | <b>ln (MHz)</b> | <b>ln (\$/#T<sup>0.07</sup>)</b> |
|-------------|-------------------|------------|-------------------|-------------------|-----------------|----------------------------------|
| Oct-85      | 386 DX            | 16         | 275,000           | 299               | 2.7726          | 4.8237                           |
| Sep-91      | 486 SX            | 25         | 1,200,000         | 349               | 3.2189          | 4.8752                           |
| Jun-91      | 486 DX            | 50         | 1,200,000         | 665               | 3.9120          | 5.5199                           |
| Aug-92      | 486 DX2           | 66         | 1,200,000         | 682               | 4.1897          | 5.5452                           |
| Mar-93      | Pentium           | 66         | 3,100,000         | 965               | 4.1897          | 5.8258                           |
| Mar-94      | Pentium           | 100        | 3,200,000         | 995               | 4.6052          | 5.8542                           |
| Mar-95      | Pentium           | 120        | 3,200,000         | 935               | 4.7875          | 5.7920                           |
| Nov-95      | Pentium Pro       | 166        | 5,500,000         | 1682              | 5.1120          | 6.3413                           |
| Nov-95      | Pentium Pro       | 200        | 5,500,000         | 1989              | 5.2983          | 6.5090                           |

SUMMARY OUTPUT

| <i>Regression Statistics</i> |        |
|------------------------------|--------|
| Multiple R                   | 0.9677 |
| R Square                     | 0.9364 |
| Adj R Square                 | 0.9273 |
| Std Error                    | 0.1541 |
| # Obs                        | 9      |

## ANOVA

|            | <i>df</i> | <i>SS</i> | <i>MS</i> | <i>F</i> | <i>Significance F</i> |
|------------|-----------|-----------|-----------|----------|-----------------------|
| Regression | 1         | 2.4470    | 2.4470    | 103.0304 | 1.939E-05             |
| Residual   | 7         | 0.1663    | 0.0238    |          |                       |
| Total      | 8         | 2.6133    |           |          |                       |

|           | <i>Coefficients</i> | <i>Standard Error</i> | <i>t Stat</i> | <i>P-value</i> | <i>Lower 95%</i> | <i>Upper 95%</i> |
|-----------|---------------------|-----------------------|---------------|----------------|------------------|------------------|
| Intercept | 2.8843              | 0.2798                | 10.3078       | 1.7512E-05     | 2.2226           | 3.5460           |
| Slope     | 0.6598              | 0.0650                | 10.1504       | 1.9388E-05     | 0.5061           | 0.8135           |

Table C-2: Data for the Calculation of Durbin-Watson Statistic

| MHz  | # Transistors | Actual Cost (\$) | Predicted Cost (\$) |
|------|---------------|------------------|---------------------|
| 16   | 275,000       | 299              | 268                 |
| 25   | 1,200,000     | 349              | 399                 |
| 50   | 1,200,000     | 665              | 630                 |
| 66   | 1,200,000     | 682              | 756                 |
| 66   | 3,100,000     | 965              | 808                 |
| 100  | 3,200,000     | 995              | 1065                |
| 120  | 3,200,000     | 935              | 1202                |
| 166  | 5,500,000     | 1682             | 1546                |
| 200  | 5,500,000     | 1989             | 1748                |
| 450  | 13.71         | -                | 4598                |
| 600  | 34.39         | -                | 4103                |
| 800  | 86.28         | -                | 5291                |
| 1000 | 216.45        | -                | 6538                |
| 1100 | 543.04        | -                | 7425                |

| $Y_i$         | $Y_{ip}$       | $e_i = Y_i - Y_{ip}$ |
|---------------|----------------|----------------------|
| Observed Cost | Predicted Cost | Estimated Error      |
| 299           | 267.81         | 31.19                |
| 349           | 398.56         | -49.56               |
| 665           | 629.66         | 35.34                |
| 682           | 756.23         | -74.23               |
| 965           | 808.18         | 156.82               |
| 995           | 1065.45        | -70.45               |
| 935           | 1201.64        | -266.64              |
| 1682          | 1546.03        | 135.97               |
| 1989          | 1748.26        | 240.74               |

| $e_i - e_{i-1}$               | $(e_i - e_{i-1})^2$ | $(e_i)^2$                 |
|-------------------------------|---------------------|---------------------------|
| Difference of Adjacent Errors | Difference Squared  | Square of Estimated Error |
| -                             | -                   | 973                       |
| -80.750                       | 6520.5              | 2456                      |
| 84.903                        | 7208.5              | 1249                      |
| -109.575                      | 12006.6             | 5511                      |
| 231.053                       | 53385.3             | 24592                     |
| -227.268                      | 51650.9             | 4963                      |
| -196.190                      | 38490.3             | 71096                     |
| 402.609                       | 162094.4            | 18488                     |
| 104.764                       | 10975.5             | 57953                     |

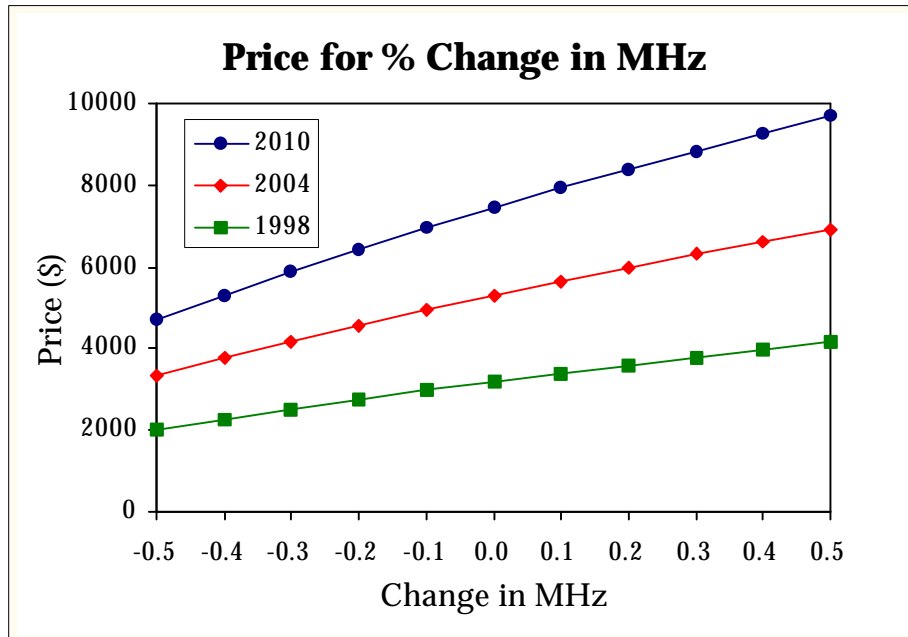
| Sum of Diff. Squared | Sum of Square of Est Error |
|----------------------|----------------------------|
| 342332.1             | 187282                     |

| Durbin-Watson Statistic |
|-------------------------|
| 1.83                    |

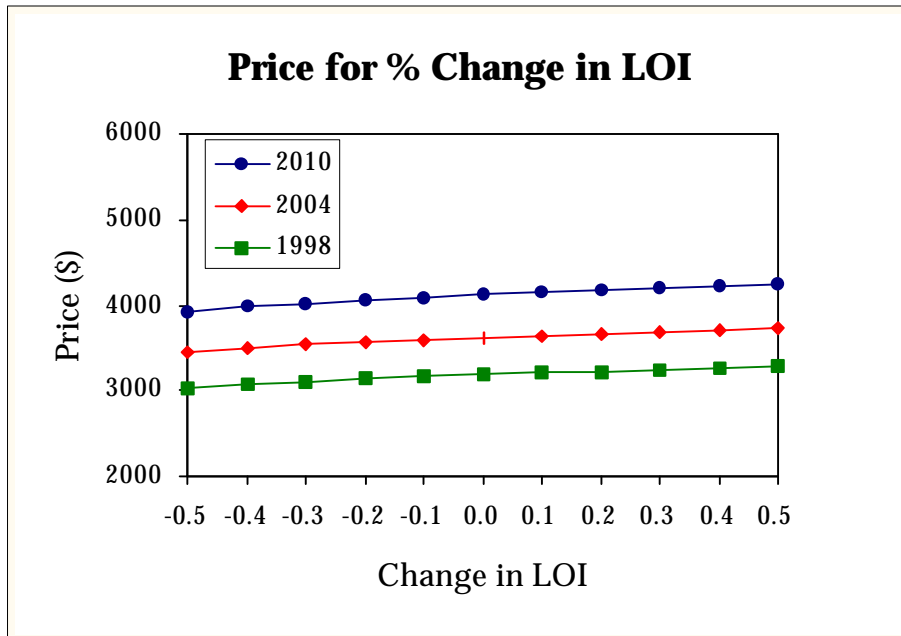
## APPENDIX D

### SENSITIVITY GRAPHS FOR TECHNICAL AND ESTIMATED PARAMETERS

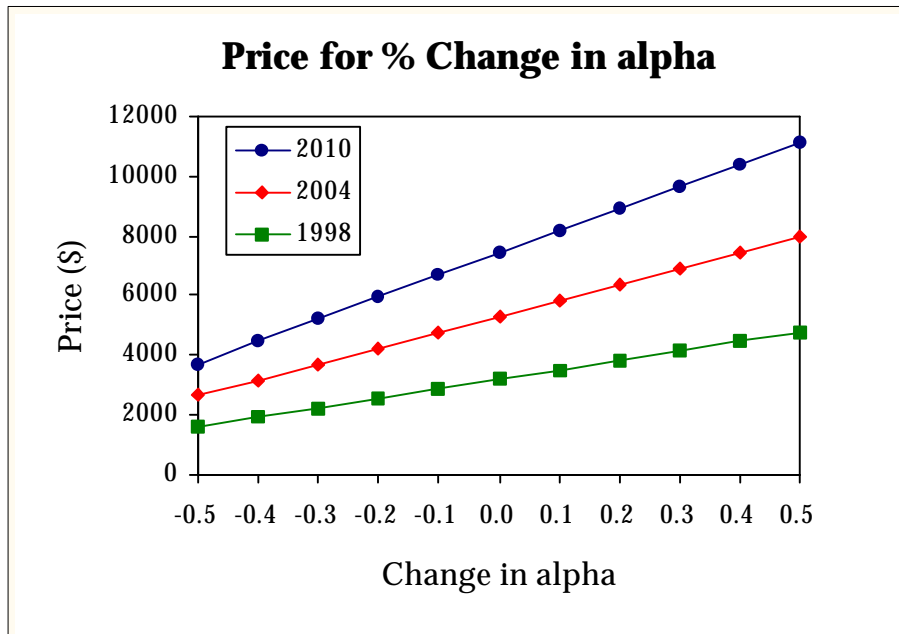
Included below are the graphs for the sensitivity analysis discussed in Chapter 5. These graphs show the change in price for % changes in the technical and estimated parameters of the Modified VHSIC Model.



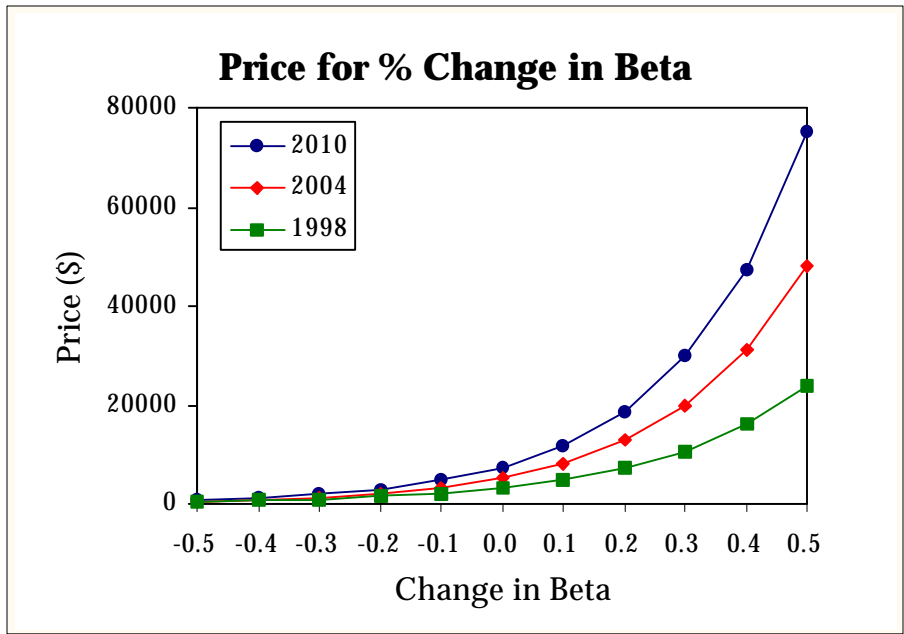
**Figure D-1:** The price change for % change in MHz is shown. Between the two technical parameters of MHz and LOI, price is more affected by MHz.



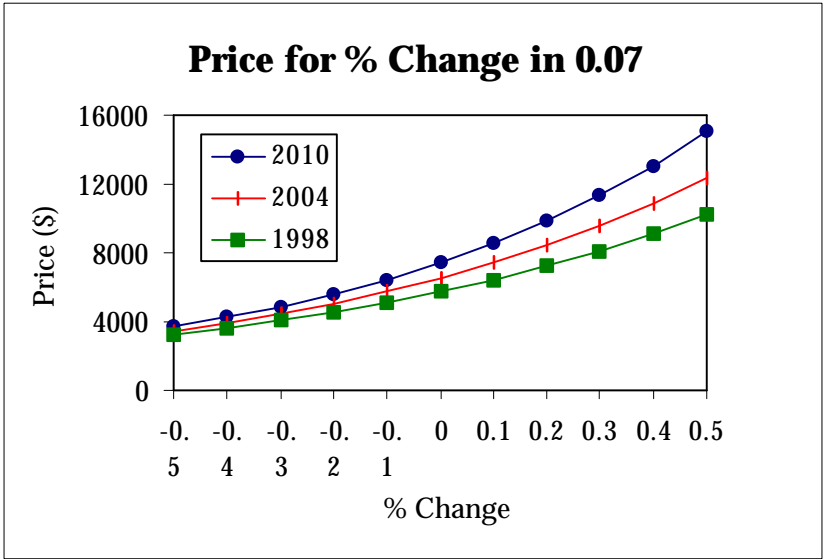
**Figure D-2:** Price change is shown for % change in LOI. Price is relatively unaffected by large changes in the LOI.



**Figure D-3:** Price is shown for % changes in  $\alpha$ . Although the initial selling price affected by changes in  $\alpha$ , price is more sensitive to changes in  $\beta$ .



**Figure D-4:** The price is shown for % change in  $\beta$ . Of the estimated parameters initial selling price is more affected by changes in  $\beta$  than  $\alpha$ .



**Figure D-5:** The price is shown for % change in the exponent of the LOI factor (0.07). Initial selling price is more sensitive to changes in the exponent of the LOI factor than changes in the LOI factor.

**APPENDIX E**

**REGRESSION STATISTICS AND GRAPHS**

**FOR PROFIT-TIME EQUATIONS**

Following are the regression statistics and graphs for the Profit-Time Equations generated in Chapter 5 of this report.

**Table E-1: MHz Data For Regression Analysis**

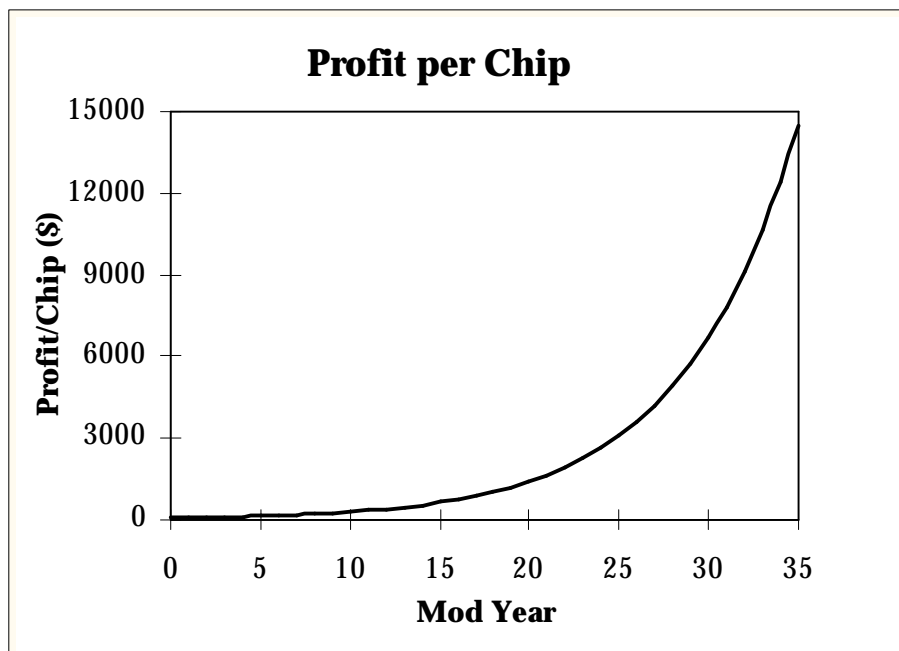
| Mod Year | MHz | ln (MHz) | Predicted MHz |
|----------|-----|----------|---------------|
| 1.5      | 5   | 1.609    | 3.742         |
| 5        | 6   | 1.792    | 7.496         |
| 8.8      | 16  | 2.773    | 15.937        |
| 12.5     | 25  | 3.219    | 33.217        |
| 16.25    | 60  | 4.094    | 69.924        |
| 18.25    | 150 | 5.011    | 104.000       |

| <b>SUMMARY OUTPUT</b> |        |
|-----------------------|--------|
| Regression Statistics |        |
| Multiple R            | 0.9784 |
| R Square              | 0.9572 |
| Adj. R Square         | 0.9465 |
| Standard Error        | 0.3049 |
| Observations          | 6      |

| ANOVA      | df | SS     | MS    | F      | Significance F |
|------------|----|--------|-------|--------|----------------|
| Regression | 1  | 8.320  | 8.32  | 89.495 | 0.000696       |
| Residual   | 4  | 0.3718 | 0.093 |        |                |
| Total      | 5  | 8.692  |       |        |                |

|            | Coefficient | SE       | t Stat | P-value  | Lower 95% | Upper 95% |
|------------|-------------|----------|--------|----------|-----------|-----------|
| Intercept  | 1.021957    | 0.250913 | 4.073  | 0.0152   | 0.3253    | 1.719     |
| X Variable | 0.19849     | 0.020982 | 9.4602 | 0.000696 | 0.1402    | 0.2567    |

1



**Figure E-1:** The profit per chip is shown as a function of time. This is a graph of Equation 5-3 derived in Chapter 5 of this report.

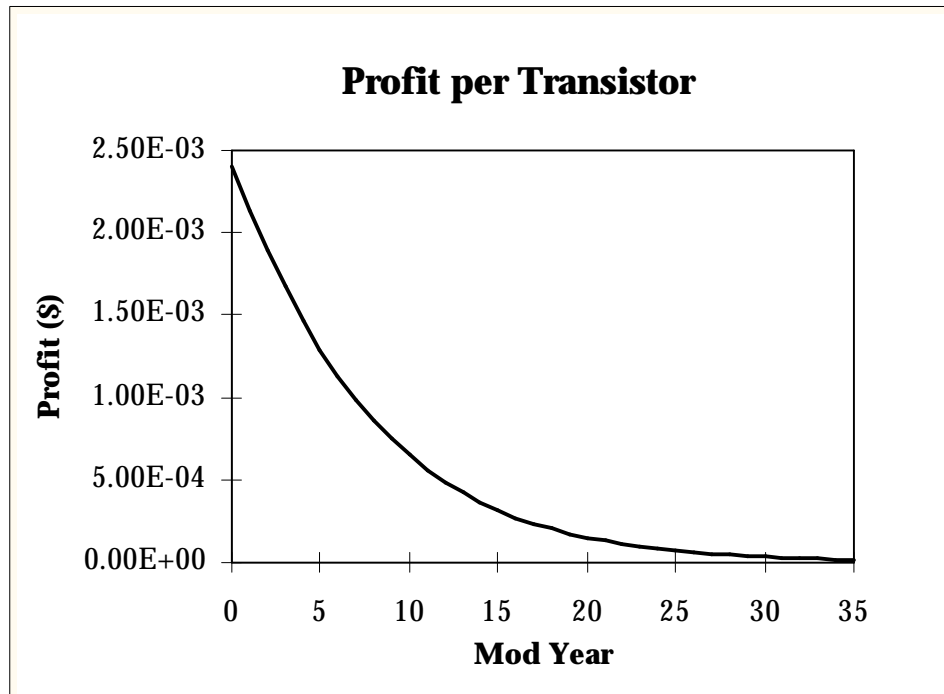
**Table E-2: Price Per Transistor Data For Regression Analysis**

| Mod Year | Price per Transistor | ln(Price) | Predicted Price |
|----------|----------------------|-----------|-----------------|
| 1.5      | 0.0031               | -5.77672  | 0.002754        |
| 5.0      | 0.00129              | -6.65442  | 0.001605        |
| 8.8      | 0.00083              | -7.09084  | 0.000893        |
| 12.5     | 0.000389             | -7.85141  | 0.000505        |
| 16.25    | 0.000238             | -8.34308  | 0.000283        |
| 18.25    | 0.000246             | -8.30883  | 0.000208        |
| 21.0     | 0.000232             | -8.36813  | 0.000136        |
| 24.0     | 0.000119             | -9.03375  | 8.56E-05        |
| 27.0     | 6.13E-05             | -9.69937  | 5.39E-05        |
| 30.0     | 3.02E-05             | -10.4076  | 3.39E-05        |
| 33.0     | 1.37E-05             | -11.2001  | 2.14E-05        |

| <b>SUMMARY OUTPUT</b> |        |
|-----------------------|--------|
| Regression Statistics |        |
| Multiple R            | 0.9841 |
| R Square              | 0.9684 |
| Adj. R Square         | 0.9649 |
| Standard Error        | 0.3016 |
| Observations          | 11     |

| ANOVA      | df | SS     | MS     | F       | Significance F |
|------------|----|--------|--------|---------|----------------|
| Regression | 1  | 25.09  | 25.088 | 275.813 | 4.65E-08       |
| Residual   | 9  | 0.8186 | 0.0909 |         |                |
| Total      | 10 | 25.907 |        |         |                |

|              | <i>Coefficients</i> | <i>SE</i> | <i>t Stat</i> | <i>P-value</i> | <i>Lower 95%</i> | <i>Upper 95%</i> |
|--------------|---------------------|-----------|---------------|----------------|------------------|------------------|
| Intercept    | -5.663              | 0.1898    | -29.84        | 2.607E-10      | -6.093           | -5.234           |
| X Variable 1 | -0.1543             | 0.00929   | -16.61        | 4.646E-08      | -0.1753          | -0.1333          |



**Figure E-2:** The profit per transistor is shown. This graph is taken from Equation 5-5 derived in Chapter 5 of this report.

## VITA

Michael D. Smith was born on June 1, 1972, in Cumberland, Maryland. From 1986 to 1990 he attended Bishop Walsh High School in Cumberland. He graduated from Virginia Polytechnic Institute and State University with a Bachelors of Science Degree in Materials Science and Engineering in May of 1994. As an undergraduate, his specialization was electronic materials.

Mr. Smith has been in the graduate program of Industrial and Systems Engineering since the Fall of 1994. His concentration in the ISE Program was Manufacturing Systems Engineering. In the Summer of 1995, he worked at the MITRE Corporation as a Technology Cost Analyst. He graduated from Virginia Tech in May of 1996 with a Master of Engineering Degree. Soon thereafter he went to work at Price Waterhouse LLP as a Systems Consultant.