

Modularity in the Design of Complex Engineering Systems

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We alone are responsible for errors, oversights and faulty reasoning.

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Introduction

In the last decade, the concept of modularity has caught the attention of engineers, management researchers and corporate strategists in a number of industries. When a product or process is “modularized,” the elements of its design are split up and assigned to modules according to a formal architecture or plan. From an engineering perspective, a modularization generally has three purposes:

- To make complexity manageable;
- To enable parallel work; and
- To accommodate future uncertainty.

Modularity accommodates uncertainty because the particular elements of a modular design may be changed after the fact and in unforeseen ways as long as the design rules are obeyed. Thus, within a modular architecture, new module designs may be substituted for older ones easily and at low cost.

This paper will make three basic points. First, we will show that *modularity is a financial force* that can change the structure of an industry. Then, we will explore the *value* and *costs* that are associated with constructing and exploiting a modular design. Finally we will examine the ways in which modularity shapes organizations and the risks that it poses for particular firms.¹

The Financial Power of Modularity

To demonstrate the financial power of modularity, let us begin by looking at some data from the computer industry. Figure 1 is a graph of the market values (in 2002 constant US dollars) of substantially all the U.S. based public corporations in the computer industry from 1950 to 2002. The firms are aggregated into sixteen subsectors by primary SIC code. The SIC codes included in the database and their definitions are listed in Table 1. IBM, Intel and Microsoft are shown separately.

¹ Some of the arguments and figures in this paper are taken from Baldwin and Clark, 2000. The figures are reprinted by permission.

Figure 1
The Market Value of the U. S. Computer Industry
By sector, 1950-2002 in constant 2002 US dollars

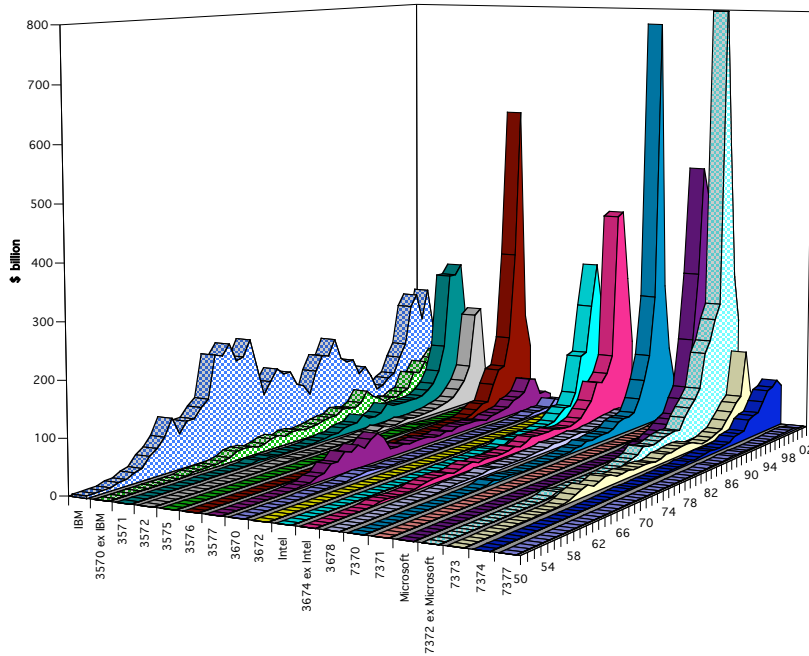


Table 1
SIC Codes Included in the Database

SIC Code	Category Definition	Start Date (1)
3570	Computer and Office Equipment	1960
3670	Electronic Components and Accessories	1960
3674	Semiconductors and Related Devices	1960
3577	Computer Peripheral Devices, n.e.c.	1962
3678	Electronic Connectors	1965
7374	Computer Processing, Data Preparation and Processing	1968
3571	Electronic Computers	1970
3575	Computer Terminals	1970
7373	Computer Integrated Systems Design	1970
3572	Computer Storage Devices	1971
7372	Prepackaged Software (2)	1973
3576	Computer Communication Equipment	1974
3672	Printed Circuit Boards	1974
7370	Computer Programming, Data Processing, and Other Services	1974
7371	Computer Programming Services	1974
7377	Computer Leasing	1974

(1) Start date is the first year in which six or more are present in the category.
 (2) This category had six firms in 1971, dipped to five in 1972, and back to six in 1973.

Figure 1 tells a story of industry evolution that runs counter to conventional wisdom. In economics the dominant theories of industry evolution describe a process of pre-emptive investment by large, well-capitalized firms, leading to stable market structures and high levels of concentration over long periods of time.² These theories are backed up by a great deal of supporting empirical evidence going back to the late 19th Century. The data in Figure 1, by contrast, show that while IBM dominated the industry in the 1950s and 1960s, in the 1970s and 1980s the computer industry “got away” from IBM. (IBM’s market value is the blue “mountain range” at the far left of the chart.) In 1969, 71% of the market value of the computer industry was tied up in IBM stock. By 2002, IBM was no longer dominant, and the largest firm (Microsoft) accounted for less than 20% of the total value of the industry.

Underlying Figure 1 is a pattern of extreme turbulence at the level of firms. The entire database spanning the years 1950 to 2002 contains about 2,700 firms. Of these, only about 1,100 survived in 2002. Thus around 1,600 or 60% of the firms that entered the computer industry over five decades no longer exist: they went bankrupt, were acquired, or moved out of the industry. Not surprisingly (for those who lived through it), much of this turnover occurred between 1997 and 2002, the years of the Internet Bubble and Crash. Around 1,200 firms entered during these six years, while 1,100 failed or were acquired.

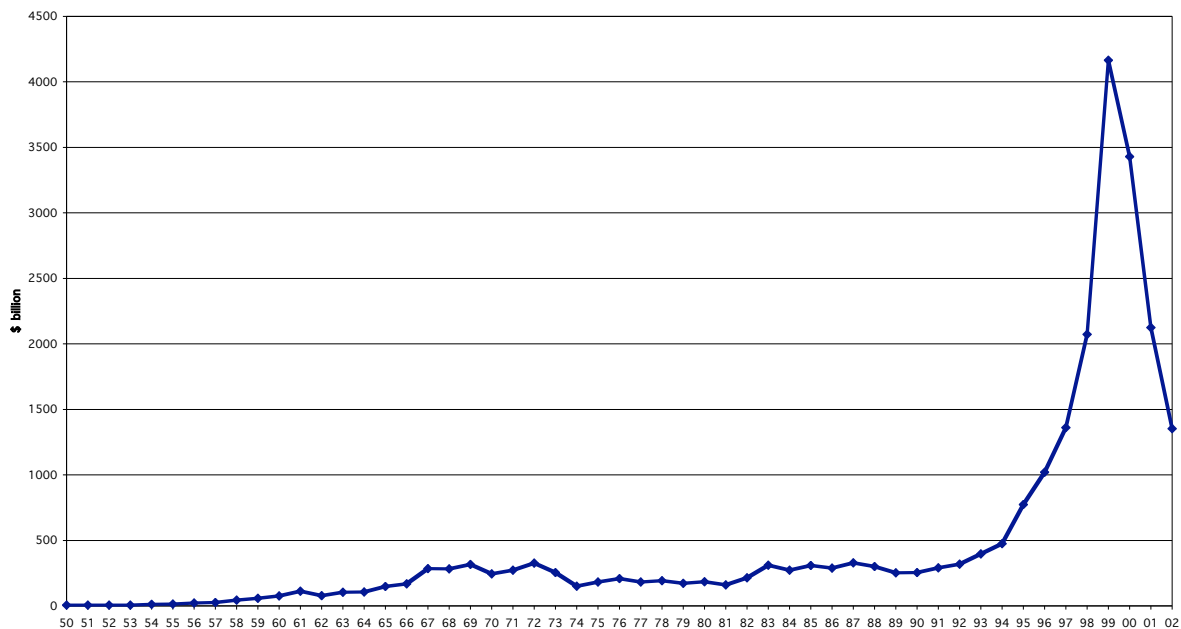
The figure also shows that market values were initially concentrated in a few firms, but are now spread out over across sixteen industrial categories. Whole industries have come and gone. For example, the original computer category, SIC 3570, “Office and Computer Equipment”, once included firms like Digital Equipment Corporation, Sperry Corporation, Data General and NCR, as well as IBM. This category has virtually disappeared: IBM has been reclassified into SIC 7370, “Computer Programming and Data Processing,” and the other firms mentioned have failed or been acquired. By 2002, Hewlett Packard was the only firm of any size remaining in this once-pre-eminent category. Conversely, in 1970,

² The original theory of pre-emptive investment leading to industry concentration, with supporting historical evidence, was put forward by Alfred Chandler (1962, 1977). A complementary theory of concentration following the emergence of a “dominant design” was put forward by William Abernathy and James Utterback (1978). Modern formulations of these theories and some large-scale empirical tests have been developed by John Sutton (1992) and Steven Klepper (1996). Oliver Williamson (1985, Ch. 11) has interpreted the structures of modern corporations (unified and multi-divisional) as responses to potential opportunism (the hazards of market contracting). It is our position that the basic “task structures” and the economic incentives of modular design (and production) systems are different from the task structures and incentives of classic large-volume, high-flow-through production and distribution systems. Therefore the organizational forms that arise to coordinate modular design (and production) may not resemble the classic structures of the modern corporation.

SIC 7372, “Packaged Software,” included only 7 firms with a combined market capitalization of just over \$1 billion. In 2002, this category had grown to 408 firms with a combined market cap of almost half a trillion dollars (\$490 billion).

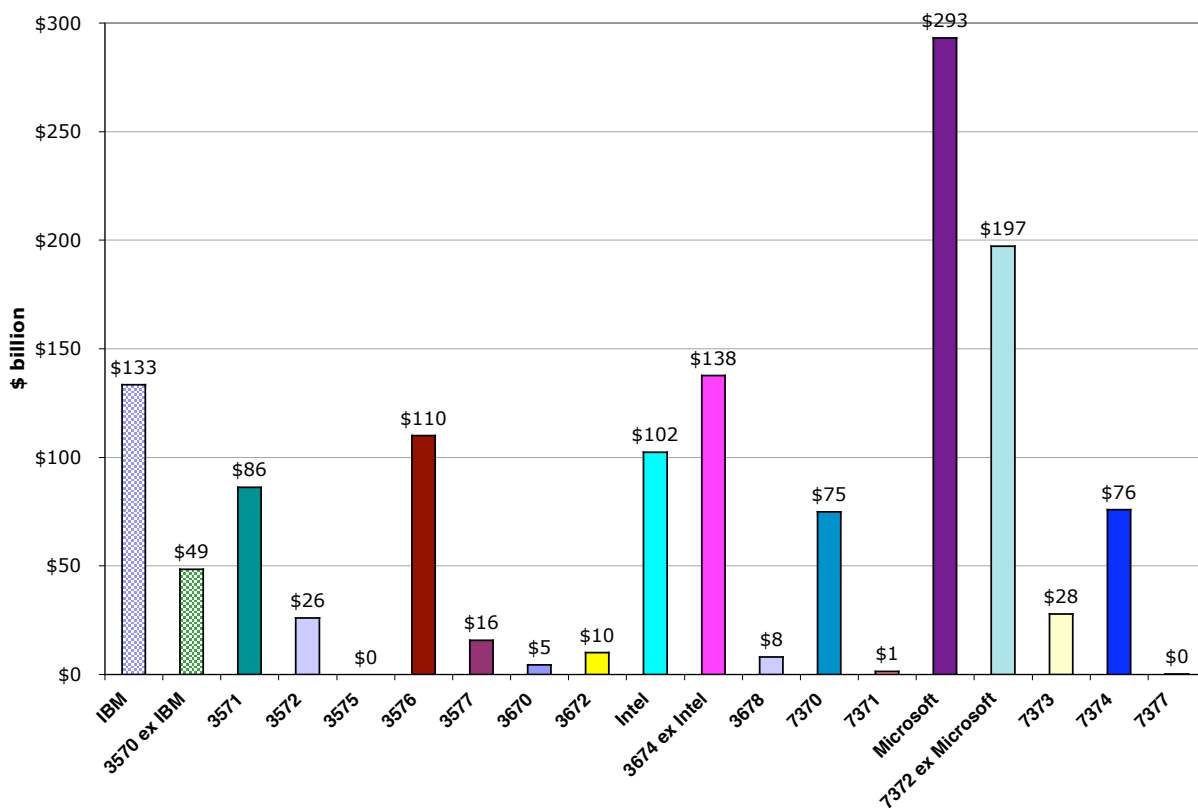
Volatility and turbulence can be observed at the level of the whole industry as well. Figure 2 shows the total market value of all firms in the industry for the sample period 1950 – 2002. The chart is dominated by the Internet Bubble and Crash, which created and then destroyed \$2.5 trillion in the space of five years (1997 – 2002). Apart from the Bubble, the industry as a whole has experienced significant value increases over time. From 1960 to 1996, even as value was being dispersed and redistributed over many, many firms, aggregate value kept pace. Then around 1997, the aggregate value of this group of firms seemed to spin out of control. More value was created and destroyed in a few years than the whole industry had managed to create over its entire history. The causes of this remarkable pattern are the subject of ongoing research, but we have no explanations for it as yet.

Figure 2
The Market Value of the Computer Industry
Aggregated, 1950-2002, in constant 2002 US dollars



In summary, the computer industry presents us with a pattern of industry evolution involving more firms, more products, and (mostly) increasing value created over time. As a result of this pattern, the computer industry today consists of a large *cluster* of over 1,000 firms, no one of which is very large relative to the whole. In addition, the total market value of the industry is now spread widely but very unevenly across the sixteen sub-industries. (See Figure 3.) We contend that modularity in the design of complex computer systems is what allowed this *creation* of value, the *dispersion* of value across so many firms, and finally new *concentrations* of value to take place. We will expand on this argument in the sections that follow.

Figure 3
The Distribution of Market Value in the U. S. Computer Industry as of 2002
By sector, in constant 2002 US dollars



The Modularity of Computer Designs

Modularity-in-design is one of the things that has caused the computer industry to evolve to its present form. In brief, modularity in computer designs was first envisioned by pioneering computer scientists like Maurice Wilkes in the 1950s. Later the goal of a modular architecture was realized by architects like Fred Brook and Gerrit Blaauw, Gordon Bell and Allen Newell, and Carver Mead and Lynn Conway in the 1960s and 1970s.³ Modular architectures in turn enabled the computer industry to evolve to its present form, which we call a “modular cluster”.

The computer industry became a cluster in approximately 1980. This cluster has been extremely dynamic, displaying high rates of entry, growth and turnover in the population of firms over time. In addition, the connections among products and companies are quite complicated in the cluster. Firms do not design or make whole computer systems; instead, they design and/or make *modules* that are parts of larger systems. These modules include hardware components like computers, microprocessors and disk drives; software components like operating systems and application programs; as well as process components like fabrication, assembly, systems integration, and testing.

Modules, in fact, are always distinct parts of a larger system. They are designed and produced independently of one another, but must function together as a whole. Modularity allows tasks—both design tasks and production tasks—to be divided among groups, which can work independently and do not have to be part of the same *firm*. Compatibility among modules is ensured by “design rules” that govern the architecture, the interfaces, and the standardized tests of the system. Thus “modularizing” a system involves specifying its *architecture*, that is, what its modules are; specifying its *interfaces*, i.e., how the modules interact; and specifying *tests* which establish that the modules will work together and how well each module performs its job.

From an engineering perspective, modularity does many things. First, it makes the complexity of the system manageable by providing an effective “division of cognitive labor.”⁴ It also makes possible the graceful evolution of knowledge about the system.⁵ In this way, modular systems are “cognitively economic.” Second, modularity organizes and enables parallel work. Work on or in modules can go on

³ See Baldwin and Clark, 2000, Chapters 6-7 on the origins of modularity in early computer designs.

⁴ O.E. Williamson, 1999, “Human Action and Economic Organization,” mimeo, University of California, Berkeley; quoted in M. Aoki, *Towards a Comparative Institutional Analysis*, 2001, MIT Press, Chapter 4.

⁵ Garud and Kumarswamy, 1995.

simultaneously, hence the start-to-finish time needed to complete the job decreases. Thus modular systems are “temporally economic.” Finally, modularity *in the design* of a complex system allows modules to be changed and improved over time without undercutting the functionality of the system as a whole. In this sense, as we indicated above, the modular design of a complex system is “tolerant of uncertainty” and “welcomes experimentation” in the modules.

Modularity in Design, Production and Use

Humans interact with artifacts in three basic ways: they design them; produce them; and use them. There are, as a result, three basic types of modularity: modularity-in-design, modularity-in-production, and modularity-in-use. We will discuss these “modularities” in reverse order.

A system of goods is *modular-in-use* if consumers can mix and match elements to come up with a final product that suits their taste and needs. For example, consumers often buy bed frames, mattresses, pillows, linens, and covers made by different manufacturers and distributed through different retailers. The parts all fit together because different manufacturers make the goods in standard sizes. These standard dimensions constitute design rules that are binding on manufacturers, wholesalers, retailers, and users. Modularity-in-use thus supports customization of the system to suit the needs and tastes of the end-user.

Manufacturers have used *modularity-in-production* for a century or more. Car makers, for example, routinely arrange to manufacture the components of an automobile at different sites and bring them together for final assembly. They can do so because they have completely and precisely specified how the parts will interact with the vehicle. The engineering specifications of a component (its dimensions, tolerances, functionality, etc.) constitute a set of design rules for the factories that supply the parts. Such process modularity is fundamental to mass production.

However, the fact that, in a complex system, the elements of use or the tasks of production have been split up and assigned to separate modules does not mean that the *design* of the system is modular. Indeed systems that are modular-in-use or modular-in-production may rest on designs that are tightly coupled and centrally controlled. For example, Intel Corporation famously imposes a “copy exactly” rule on its fabrication plants. The production of chips can go on independently at separate sites because the layout of the plants and the work processes within the plants are the same. Thus Intel’s “copy exactly”

plants are modular-in-production but not modular-in-design. In a similar vein, a sectional sofa is a suite of furniture that is modular-in-use. Purchasers can combine and recombine the elements of the suite at will. But those elements must be designed as one interdependent whole, or the patterns and shapes will not form a pleasing ensemble. Thus the sectional sofa suite is modular-in-use, but not modular-in-design.

A complex engineering system is modular-in-design if (and only if) the *process of designing it can be split up and distributed across separate modules*, that are coordinated by design rules, not by ongoing consultations amongst the designers. Of all the “modularities”, modularity-in-design is the least well understood and has the most interesting economic consequences. This is because new designs are fundamentally *options* with associated economic *option value*. Modularity-in-design multiplies the options inherent in a complex system. This in turn both increases the total economic value of the system and changes the ways in which the system can evolve. In the rest of this paper, we will explain how to map and measure the option value of modularity-in-design.

Designs as Options

A fundamental property of designs is that at the start of any design process, the final outcome is uncertain. Once the full design has been specified and is certain, *then the development process for that design is over*.

Uncertainty about the final design translates into uncertainty about the design’s eventual value. How well will the end-product of the design process perform its intended functions? And what will it be worth to users? These questions can never be answered with certainty at the beginning of any substantive development process. Thus the ultimate value of a design is unknown when the development process begins.

Uncertainty about final value in turn causes new designs to have “option-like” properties. In finance, an option is “the right but not the obligation” to choose a course of action and obtain an associated payoff. In engineering, a new design creates the ability but not the necessity—the right but not the obligation—to do something in a different way. In general (if the designers are rational), the new design will be adopted only if it is better than its alternatives. Thus the economic value of a new design is properly modeled as an option using the methods of modern finance theory.

The option-like structure of designs has three important but counterintuitive consequences. In the first place, when payoffs take the form of options, taking more risk creates more value.⁶ Risk here is defined as the *ex ante* dispersion of potential outcomes. Intuitively, a risky design is one with high technical potential but no guarantee of success. “Taking more risk” means accepting the prospect of a greater *ex ante* dispersion. Thus a risky design process is one that has a very high potential value conditional on success but, symmetrically, a very low, perhaps negative, value conditional on failure.

What makes the design an option, however, is that the low-valued outcomes do not have to be passively accepted. As we said, the new design does not have to be adopted; rationally, it will be adopted only if it is better than the alternatives, including the *status quo* alternative. In effect, then, the downside potential of a risky design is limited by the option to reject it after the fact. This means that “risk” creates only upside potential. More risk, in turn, means more upside potential, hence more value.⁷

The second counterintuitive result is that when payoffs take the form of options, seemingly redundant efforts may be value-increasing. Two attempts to create a new design may arrive at different endpoints. In that case, the designers will have the option to take the better of the two. The option to take the better of two or best of several outcomes is valuable.⁸ Thus when faced with a risky design process, which has a wide range of potential outcomes, it is often desirable to run multiple “design experiments” with the same functional goal. These experiments may take place in parallel or in sequence, or in a combination of both modes.⁹ But whatever the mode, more risk calls for more experimentation.

The third result is that *options interact with modularity in a powerful way*. By definition, a modular architecture allows module designs to be changed and improved over time without undercutting the functionality of the system as a whole. This is what it means to be “tolerant of uncertainty” and to “welcome experiments” in the design of modules. As a result, modules and design experiments are

⁶ This is a basic property of options. See Merton, 1973.

⁷ It follows, of course, that if a risky design is “hardwired” into a system so that it must be implemented regardless of its value, then the design process loses its option-like properties. In such cases, “taking more risk” in the sense defined above, will not increase, and may decrease value.

⁸ Stulz, 1982, first analyzed the option to take the higher-valued of two risky assets. Sanchez (1991) worked out the real option value of parallel design effort in product development.

⁹ Loch, Terwiesch and Thomke, 2001.

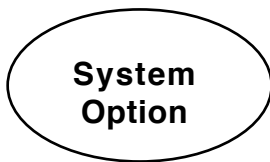
economic complements: an increase in one makes the other more valuable.¹⁰ (Below we will derive this result in the context of a formal model.)

The effect of modularity-in-design on options and option value is depicted in Figure 4. Here we envision a system that is making the transition from being one interdependent whole to being a set of modules governed by design rules. The system goes from having one large design option (i.e., to take the whole design or leave it) to having many smaller options—one per module. Thus the act of *splitting* a complex engineering system into modules multiplies the valuable design options in the system. At the same time, this *modularization* moves decisions from a central point of control to the individual modules. The newly decentralized system can then evolve in new ways.

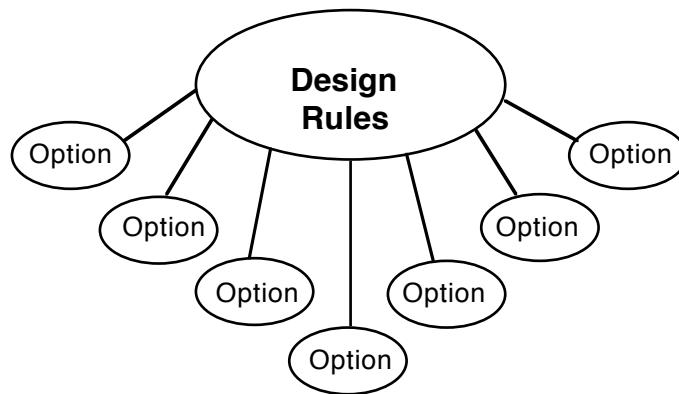
Notice, however, that by modularizing, one barrier to entry by competitors, the high costs of developing an entire complex engineering system (like an automobile, a computer, or a large software package) are reduced to the costs of developing individual modules. Thus the modularization of a large, complex system, even as it creates options and option value, also sows the seeds of increased competition focused on the modules. We shall revisit this issue at the end of the paper.

Figure 4
Modularity Creates Design Options

System Before Modularization



System after Modularization



¹⁰ This is the definition of economic complementarity used by Milgrom and Roberts (1990) and Topkis (1998, p. 43). The complementarity of modularity and experimentation was first demonstrated by Baldwin and Clark (1992; 2000, Chapter 10).

Mapping the Design of a Complex Engineering System

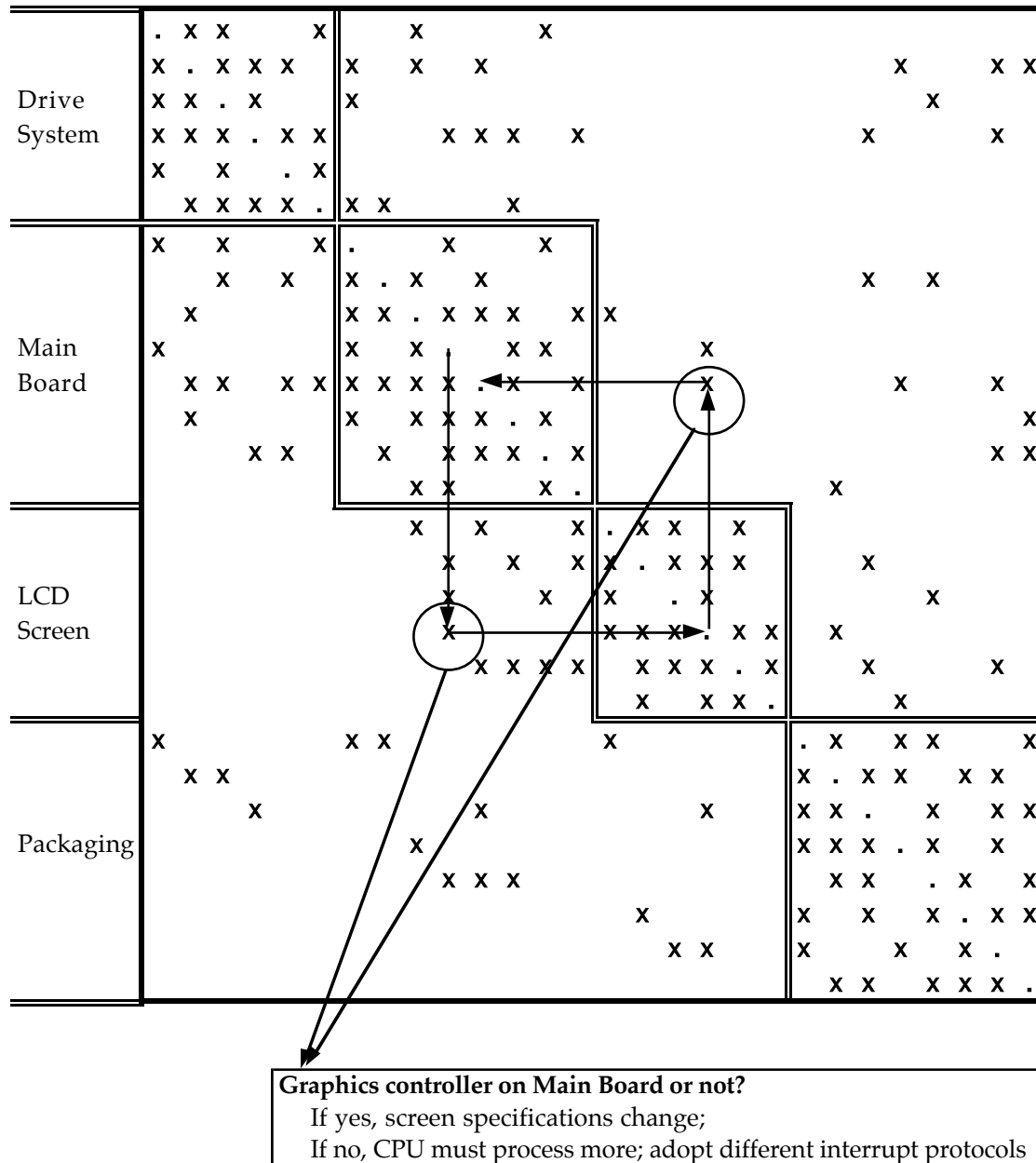
We will now look at modularity-in-design more carefully. To help us do so, we will represent the design of a complex system using the methods of Design Structure Matrix (DSM) Mapping. In this mapping technique, the system is first characterized by listing a set of design parameters for the system. The design parameters are then arrayed along the rows and columns of a square matrix. The matrix is filled in by checking—for *each parameter*—which other parameters affect it and which are affected by it. For example, if Parameter A affects the choice of Parameter B, then we will put a mark “**X**” in the cell where the *column of A* and the *row of B* intersect. We repeat this process until we have recorded all parameter interactions. The result is a map of the dependencies that affect the detailed structure of the artifact. For example, Figure 5 is a DSM map of the dependencies in the design for a laptop computer system circa 1993.¹¹

DSM maps are well known in the engineering professions. They can be constructed for any artifact or complex system, whether it is tangible or intangible. Thus there are DSM maps of products, like computers and automobiles, and DSM maps of both production processes and design processes. Many such maps have been constructed by Steven Eppinger and his colleagues at MIT.

The DSM map in Figure 5 indicates that the laptop computer design has four blocks of very tightly interrelated design parameters corresponding to the (Disk) Drive System, the Main Board, the LCD Screen, and the Packaging of the machine. There is also a scattering of dependencies (“**X**’s”) outside the blocks. The dependencies arise both above and below the main diagonal blocks, thus the blocks are *interdependent*.

¹¹ The DSM methodology was invented by Donald Steward. The DSM map shown in Figure 5 was prepared by Kent McCord and published in McCord and Eppinger, 1993. Reprinted by permission.

Figure 5
Design Structure Matrix Map of a Laptop Computer



Cycling and iteration are needed to resolve design interdependencies. For example, as shown in the figure, the location of the computer's graphics controller creates dependencies between the Main Board and the LCD Screen and vice versa. Because of these dependencies, there will be ramifications of

any choice made at this particular point: these are indicated by the arrows in the diagram. If two teams were working on the different components, they would have to confer about the location of the graphics controller in order to coordinate their design choices. But unforeseen consequences might arise later, causing the initial choice to be revisited. There would then be further consequences: new arrows would arise, which, through the chain of dependencies, might wander all over the map. Such cycling is the inevitable consequence of an interdependent design structure.

However, it is important to note that the DSM map for a product or process need not be set in stone forever. Dependencies and interdependencies can be modified by a process of design rationalization, which works in the following way. Suppose that the designers of the laptop computer system wished to eliminate the interdependencies between the Main Board and the Screen that were due to the graphics controller location. They could do so by setting a *design rule* that located the graphics controller on the Board (for example). By this action the two design teams would have restricted their alternatives, *but* they would also have eliminated a source of cycling between two of the blocks of the design.

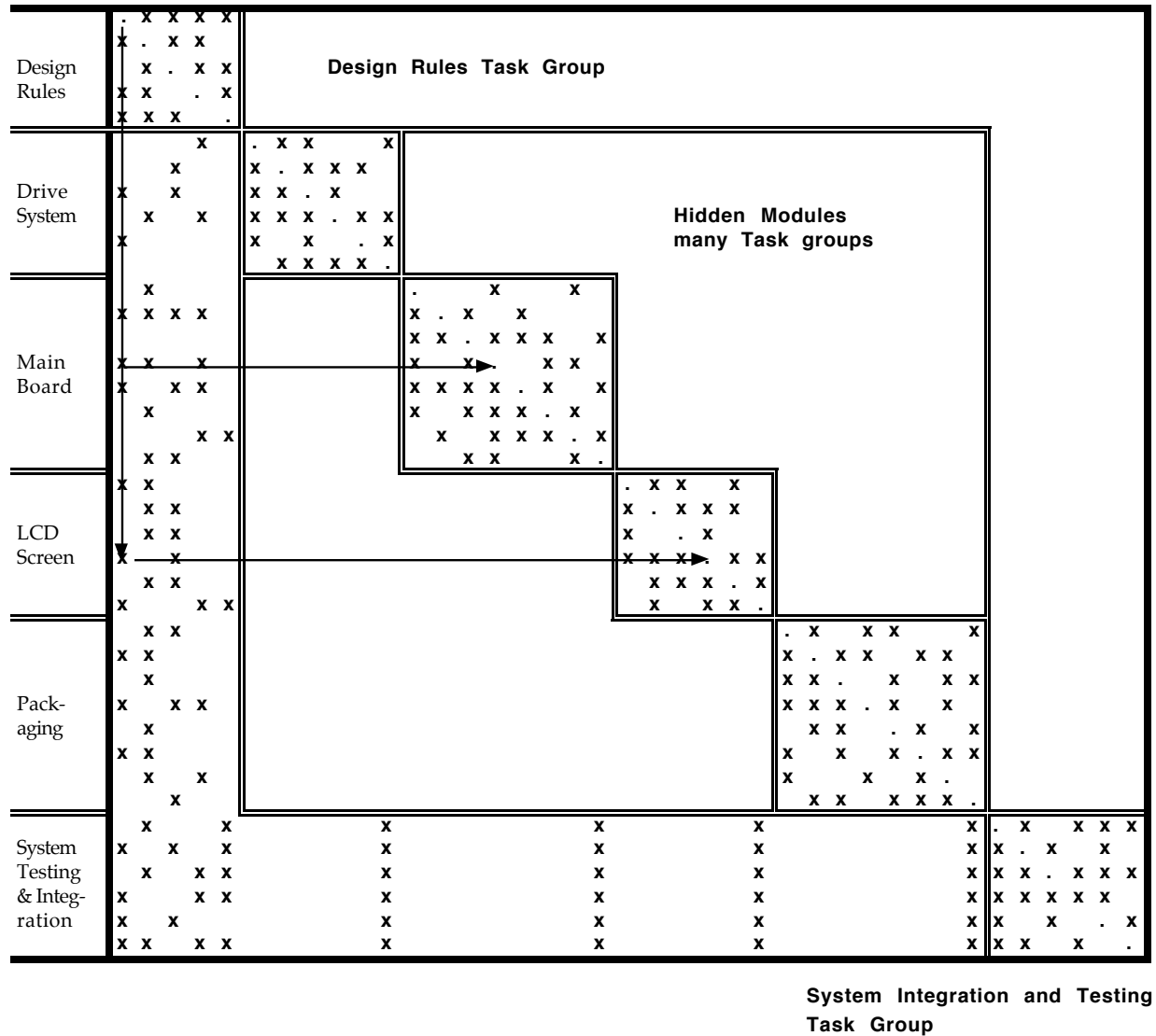
Figure 6 shows the new DSM map obtained by turning the graphics control location into a design rule. Two dependencies, one above the diagonal and one below, which were present before, now do not exist: they are absent from the circled areas in the map. Instead there is a design rule that is known (hence “visible”) to both sets of designers, which they must obey.

Figure 6
Eliminating Interdependencies by Creating a Design Rule

Design Rules	Graphics Controller - Yes/no										
Drive System											
		.	x	x		x					
		x	.	x	x	x					
		x	x	.	x						
		x	x	x	.	x	x				
Main Board		x	x	x							
			x	x							
		x									
	x	x	x	x	x	x	x				
	x	x	x	x	x	x	x				
LCD Screen											
Packaging											

Carrying this process through to its logical conclusion results in a radically different structure: a *modular structure* as shown in Figure 7. Here we have the same highly interdependent blocks as before: the Drive System, the Main Board, the LCD Screen, and Packaging. And within those blocks essentially nothing has changed, the pattern of interdependency is the same. *But the out-of-block dependencies both above and below the main diagonal have all disappeared.*

Figure 7
Modularization of a Laptop Computer Design



How does that happen? First, in the new structure, each of the former out-of-block dependencies has been addressed by a design rule. Thus, there is now a new “Design Rules” block (not drawn to scale), whose parameters affect many of the parameters in the component blocks. Those dependencies are indicated by the “X”s in the vertical column below the Design Rules block. (Design rule parameters are often called “standards.”)

By obeying the design rules, teams working on the designs of each of the component blocks—which are now modules—can maintain conformity with the other parts of the system. But note that there has been another, earlier stage in the process in which the design rules were established.

Furthermore, the new process, as shown in Figure 7, delivers *four separate items*, which must still be integrated into a functioning whole system. No set of design rules is perfect, and unforeseen compatibility problems are often revealed in the latter stages of a modular design process. For these reasons, a “System Integration and Testing” (SIT) block appears in the lower right corner of the modular DSM. This block is affected by the design rules and by some parameters of the hidden modules. But decisions taken in the SIT block, by definition, will not affect choices in the prior blocks. (If they do, then the structure is no longer modular).

Therefore, a modular design structure has three characteristic parts:

- *design rules*, which are known and obeyed by teams responsible for individual modules;
- so-called *hidden modules* that “look to” the design rules, but are independent of one another as work is proceeding; and
- a *systems integration and testing module* in which the hidden modules are assembled into a system, and any remaining, minor problems of incompatibility are resolved.

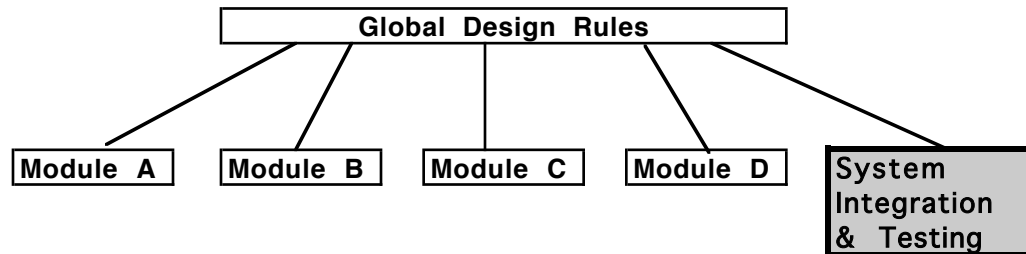
A complex system design may go from being *interdependent* to being *modular* in the following way. The “architects” of the system must first identify the dependencies between the distinct components and address them via a set of design rules. Second, they must create encapsulated or “hidden” modules corresponding to the components of the system. And third, they must establish a separate system integration and testing activity that will assemble the modular components and resolve unforeseen incompatibilities.

The Design Hierarchy Representation

A DSM map is one way to represent a modular system: a design hierarchy is another.¹² A design hierarchy shows which modules are affected by which other modules. (See Figure 8.) At the very top of the design hierarchy are the system-wide design rules: these must be obeyed (hence “are visible to”) all modules in the system.

¹²Marples 1961; Clark, 1985.

Figure 8
A Two-level Modular Design Hierarchy



The system comprises four "hidden" modules and a "System Integration and Testing" stage. From the standpoint of the designers of A, B, C, and D (the hidden modules), system integration and testing is simply another module: they do not need to know the details of what goes on in that phase as long as they adhere to the global design rules. Obviously, the converse is not true: the integrators and testers have to know something about the hidden modules in order to do their work. How much they need to know depends on the completeness of the design rules. Over time, as knowledge accumulates, more testing will occur within the modules, and the amount of information passed to the integrators will decrease. The special, time-dependent role of integration and testing is noted by the heavy black border around and gray shading within the "module."

Below the system-wide design rules, there may be "architectural modules," which establish design rules for certain subsystems. There are no architectural modules in Figure 8, but in most complex engineering systems there are one or two layers of architectural modules. For example, operating systems like Microsoft Windows and Unix are architectural modules in a computer system.

Finally, at the bottom of the design hierarchy are the hidden modules of the system: these must obey the design rules, hence "look to" them. But the hidden modules' own parameters are "encapsulated": they do not affect, and hence do not need to be known to those working on other modules. Hidden modules are thus the primary source of option value in a modular system. (Even so, depending on the rules governing intellectual property, much of the value created in the hidden modules may be captured by the companies that control architectural modules and/or design rules.)

Modular Organizations

The “modularized” architectures depicted in Figures 7 and 8 lead naturally to a “modularized” organizational structures.¹³ In a recent paper which addressed the nature of transactions we postulated that the activities represented by the x’s in Figures 5 and 6 naturally map onto organizations since each interaction captured on the DSM represents a transfer of material or information or both.¹⁴ In this way it is natural to look at Figure 6 and see a ‘traditional’ organization structured around specific technologies or disciplines. The within-block interactions represent interactions that are internal to each organizational unit while the out-of-block interactions can be viewed as interactions that require coordination across units. In this system, each organizational unit would have liaison personnel whose function was to assure that activities in each unit of the overall endeavor remained synchronized and coordinated.

In the modularized design of Figures 7 and 8 however, many of the cross-unit liaison and coordination functions have been eliminated. This is done in two ways: through design rules and through encapsulation. Design rules (standards) ensure that decisions that affect multiple units are fixed and communicated ahead of time, and not changed along the way. Encapsulation means that all activities associated with specific aspects of the product are conducted within one organizational unit – even if that organizational unit embodies multiple skillsets, disciplines or activities. This may lead to duplication of skillsets – for example the Main Board and Packaging groups may both need people who are skilled in understanding interconnections – but, if they are to be encapsulated, these groups must be scaled and staffed to their own needs without calling on shared resources, and without requiring extensive cross-unit coordination activities.

Similarly note that the transfers of material and information between groups have been simplified to a simple handoff from each module task group to a final systems integration and testing group. If these handoffs as well as the design rules can be standardized and codified, then there is no need for the various groups to reside in the same company: armslength transactions between several firms can cost-effectively replace complex coordinating flows of information within a single firm.

¹³ Sanchez and Mahoney, 1996.

¹⁴ Baldwin and Clark, 2002.

An educational analogy might be useful here. Instead of a laptop computer, Figure 6 might as easily represent the traditional departmental or discipline-based organizational structure of a university (like Harvard). A discipline-based organizational structure is well-suited to teaching courses, but it is ill-suited to carrying out broadly-based research initiatives that cut across many disciplines. Indeed, in order to conduct interdisciplinary research, the traditional departmental structure requires many cross-unit interactions as shown in Figure 6. These interactions are generally time-consuming and prone to cycling. In addition, many task-relevant interactions may get lost in the shuffle and not take place at all.

By contrast, a center-based or project-based collaborative structure gathers participants from multiple disciplines and organizes them into self-contained teams as in Figure 7. Even though some interactions are lost, and there may be duplication of resources across centers, the new organizational structure imposes a much smaller coordination burden on the overall research endeavor. The reduced coordination costs in turn can offset the opportunity losses and costs of implementing the more modular, team-based structure. If the coordination cost savings are large, then a “more modular” organization is a good idea. But there is always a tradeoff—some things are lost while other things are gained.

Modular Operators

A key benefit of systems with modular designs is that, especially at the lower levels of the design hierarchy, such systems can evolve. The information that is encapsulated in the hidden modules can change, as long as the design rules are obeyed. Therefore, as we said earlier, modular systems are “tolerant of uncertainty” and “welcome experimentation” *in their modules*. In contrast, the design rules of a modular system, once established, tend to be both rigid and long lasting.

There are certain generic design actions one can apply to a modular system. Following the lead of John Holland of the University of Michigan,¹⁵ we have labeled such actions “operators.” In our prior work, we identified six modular operators, and analyzed the sources of their economic value. In particular, given a modular structure, one can:

- *split* any module;
- *substitute* a newer module design for an older one;

¹⁵ Holland, 1992.

- *exclude* a module;
- *augment* the system by adding a module that was not there before;
- collect common elements across several modules and organize them as a new level in the hierarchy (modular *inversion*); and
- create a “shell” around a module so that it works in systems other than one for which it was initially designed (modular *porting*).

Figure 9 shows how each of these operators affects the structure of a modular system.

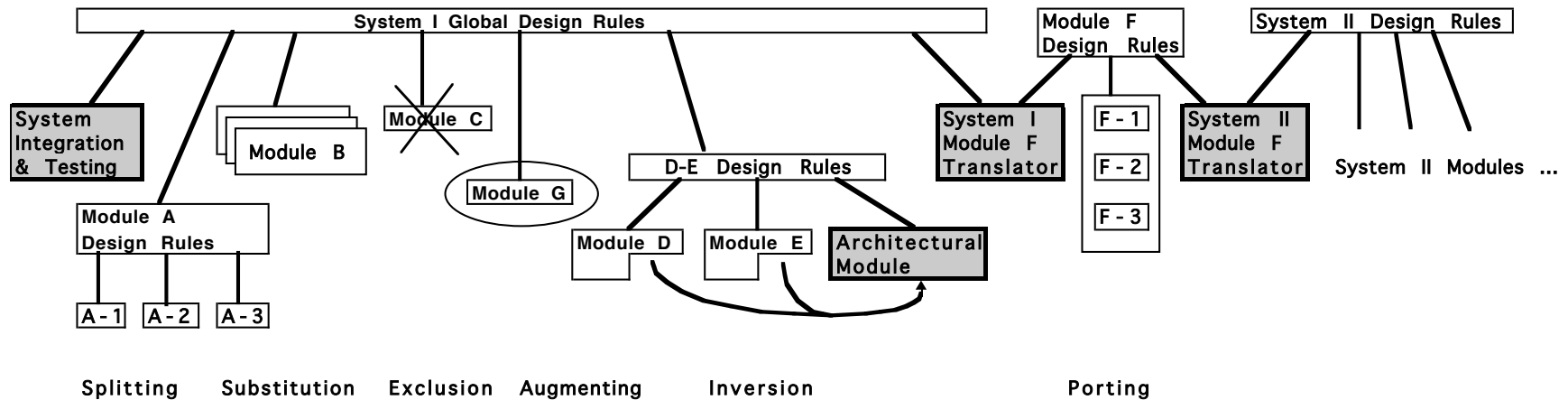
We must emphasize that we regard our list of six operators as the beginning of a useful taxonomy. The list is by no means exhaustive. Indeed three other operators have been identified in empirical investigations of design evolution. These are:

- the *linking* of two pre-existing modules;¹⁶
- the *recombining* of two previously separate modules (this is the opposite of splitting); and
- *embracing and extending* a pre-existing module (this operator was famously used by Microsoft on Sun’s version of Java).

The important thing to understand is that *operators correspond to search paths in the design space of a complex engineering system*. These search paths in turn are *options* in the so-called “value landscape” of the complex system. As options, the operator / search paths can be valued using fairly standard analytic techniques from finance. Thus, for example, the decision to *split* a complex system (or subsystem) into several modules can be valued. The decision to *augment* the system by designing several variants of a module customized for different users or purchasers can also be valued. In the next section, we will describe the economic structure of option values for the modules of a complex engineering system.

¹⁶ Bala Iyer, 2003, private communication.

Figure 9
The Effect of the Six Operators on a Modular System



We started with a generic two-level modular design structure, as shown in Figure 5-3, but with six modules (A, B, C, D, E, F) instead of four. (To display the porting operator, we moved the "System Integration & Testing Module" to the left-hand side of the figure.) We then applied each operator to a different set of modules.

Module A was **Split** into three sub-modules.

Three different **Substitutes** were developed for Module B.

Module C was **Excluded**.

A new Module G was created to **Augment** the system.

Common elements of Modules D and E were **Inverted**. Subsystem design rules and an architectural module were developed to allow the inversion.

Module F was **Ported**. First it was split; then its "interior" modules were grouped within a shell; then translator modules were developed.

The ending system is a three-level system, with two **modular subsystems** performing the functions of Modules A, D and E in the old system. In addition to the standard hidden modules, there are three kinds of special modules, which are indicated by heavy black borders and shaded interiors:

- System Integration & Testing Module
- Architectural Module
- Translator Module(s)

Option Values in a Complex Engineering System

We begin this section by introducing some notation. We assume that the total economic value of a complex engineering system can be expressed as the sum of a minimal system value, S_0 , plus the incremental value added by the performance of each of J modules. Equation 1 thus denotes the *ex post* value that will be realized once the system's design is complete:

$$\text{Economic Value of the System} = S_0 + \sum_{j=1}^J X_j^1 ; \quad (1)$$

At the beginning of our analysis, we assume that the minimal system exists and its economic value is known. Without loss of generality, we normalize that value to zero. At the same time, we assume that the modules of the system have not yet been realized, hence their eventual economic payoffs (the X_j^1) are uncertain.

As long as economic payoffs can be expressed in terms of money (e.g., a present value), module values can be modeled as one-dimensional random variables. We use superscripts to denote the *realizations* of random variables, and subscripts to denote the *distributions* of random variables with different distributions. Thus X_j^1 should be read as "the economic value of a single realization of the random variable X_j ". Total system value is a sum of realizations over a set of J random variables, with different distributions, indexed by j .

The realization of a module design is the outcome of a development effort targeted at that module. The realization in turn can have positive or negative value. A design with negative value is not worth incorporating into the system: it subtracts more functionality than it adds. At the end of a design interval, the developers can observe the realization for each module and compare that value to zero. If the new module design has positive value, it will be added to the system, and the system's value will increase by that amount. If the new module design has zero or negative value, it can be discarded, and the developers can try again. In this fashion, the developers can mix and match old and new module designs. The ability to accept or reject a particular realization is the developers' basic option and the focus of our analysis.

For simplicity, we assume that the firm or firms developing the complex system are risk-neutral expected-value maximizers, and that design intervals are short enough that we can ignore the time value

of money. In that case, the *ex ante* economic value of the entire system (whose *ex post* value is given by equation 1) can be expressed as follows:

$$V_J = S_0 + \text{Emax}(X_1^1, 0) + \text{Emax}(X_2^1, 0) + \dots + \text{Emax}(X_J^1, 0) \quad ; \quad (2)$$

Equation 2 indicates that each module's realized value will be compared to a benchmark equal to zero. If the new module design has value greater than the benchmark, the new design will be incorporated into the system, otherwise it will be rejected. Thus the expectation of the value of the new design is the *maximum* of its realization and zero. The expectation of the maximum of a random variable and a scalar is larger than the expectation of the random variable alone, thus the option to reject module designs adds to the economic value of the system.

Equation 2 is very general. We can gain further insight by specializing the assumptions. For example, as a thought experiment, consider a system with a total of N design parameters, and think of allocating the parameters into J distinct modules of different sizes. Let X_{\square} denote the economic value of a module of size $\square N$ where \square is less than (or equal to) one and the set of \square s sums to one:

$$\sum_{j=1}^J \square_j = 1 \quad .$$

For purposes of illustration, assume that X_{\square} is a normally distributed random variable with mean zero and variance $\square^2 N$: $X_{\square} \sim N(0, \square^2 N)$. In this case, the *variance* of a module's value will be proportional to the number of design parameters in the module. Roughly speaking, the dispersion of outcomes increases as a module's "complexity" measured by $\square N$ goes up.

Define z_{\square} as:

$$z_{\square} = \frac{X_{\square}}{\square (\square N)^{1/2}} \quad .$$

z_{\square} is a standard normal variant with mean zero and variance one: $z_{\square} \sim N(0, 1)$.

Substituting for the X s in terms of z in equation 2, suppressing S_0 , and collecting terms, we have:

$$V_{\square} = V(\square_1, \dots, \square_J; \square, N) = \square N^{1/2} (\square_1^{1/2} + \square_2^{1/2} + \dots + \square_J^{1/2}) \text{Emax}(z, 0) \quad . \quad (3)$$

Here $\text{Emax}(z, 0)$ is the expectation of the right half of a truncated standard normal distribution and equals .3989. Note that the system value depends on the *elements* of the vector of $(\square_1, \dots, \square_J)$ as well as the

system parameters α and N . This underscores the fact that impact of modularity cannot be captured by a single summary measure or statistic (e.g., the average degree of modularity). The *details* of the modular structure (i.e., the elements of the vector α) affect the system's option value in important and nonlinear ways.

Several results follow directly from equation 3. For example, we can compare the value of a modularized system to the value of the corresponding unmodularized system:

Proposition 1. Under the assumptions given above, let an engineering system of complexity N be partitioned into J independent modules of complexity $(\alpha_1 N, \alpha_2 N, \dots, \alpha_J N)$ respectively. The modularized system has value:

$$V_{\alpha} = (\alpha_1^{1/2} + \alpha_2^{1/2} + \dots + \alpha_J^{1/2}) V_1 ; \quad (4)$$

relative to V_1 , the value of the corresponding unmodularized system.

Proof.

By definition, a one-module design has both J and α_j equal to one. Thus $V_1 = \alpha N^{1/2} E_{\max}(z, 0)$. Collecting terms and substituting in equation (3) yields the result. QED.

From the fact that $\langle \alpha_1, \alpha_2, \dots, \alpha_J \rangle$ are fractions that sum to one, it follows that the sum of their square roots is greater than one. Thus, as expected, under these very specialized assumptions, a modular design is "always" more valuable than the corresponding non-modular design. Moreover, additional modularization (the splitting) increases value: if a module of size α is split into sub-modules of size α and β such that $\alpha + \beta = \alpha$, then the two modules' contribution to overall value will rise because $\alpha^{1/2} + \beta^{1/2} > \alpha^{1/2}$.

In this fashion, higher degrees of modularity can increase the value of a complex design *through option value*. This result is a special case of a well-known theorem, first stated by Robert Merton in 1973.¹⁷ For general probability distributions, assuming aggregate value is conserved, Merton showed that a "portfolio of options" is more valuable than an "option on a portfolio."

Up to this point in our thought experiment, we have assumed that designers will create only *one* new design per module. However, as we indicated above, an important fact about options is that

“duplication of effort,” in the sense of mounting several design experiments aimed at the same target, may be desirable. Pursuing several experimental designs gives developers the opportunity to select the best outcome after the fact. How much economic value does this option create? Is it worth the cost? The answer to this question, it turns out, depends on both the modular structure of the overall system and the technical potential inherent in each module.

To quantify the value of parallel experimentation, let us suppose that in each of J modules, the designers initiate k_j independent design efforts. When the designs are complete, the designers then have the *option* in each module to select the best of the k_j outcomes for the final design. Thus let $Q(k_j)$ denote the expected value of the highest realization of k independent draws from a standard normal distribution as long as the realization is greater than zero. Formally:

$$Q(k) = k \int_0^{\infty} z [N(z)]^{k-1} n(z) dz$$

where $N(z)$ and $n(z)$ are respectively the standard normal distribution and density functions.¹⁸

By similar reasoning as above, the total value of a system with J modules and k_j experiments on the j th module is:

$$V(\underline{\sigma}, \underline{k}; \underline{\sigma}, N) = \sigma N^{1/2} \prod_{j=1}^J \sigma_j^{1/2} Q(k_j) = \prod_{j=1}^J \sigma_j Q(k_j) . \quad (5)$$

The rightmost expression simply notes that the option value of each module is the product of the module’s dispersion parameter, σ_j , times a “highest draw expectation” for a standard normal variate. We have tabulated $Q(k)$ for values of k up to 50: the results are shown in Table 2. Using the tabulated values it is straightforward to operationalize this valuation methodology.

¹⁷ Merton, 1973.

¹⁸ This distribution of the best of k realizations is well known in statistics: it is the distribution of the “maximum order statistic of a sample of size k .” Our expectation differs from the standard one, however, because it is taken only over the range of values above zero. See Lindgren, 1968, on order statistics in general.

Table 2
Tabulated Values of Q(k) for k = 1, ..., 50

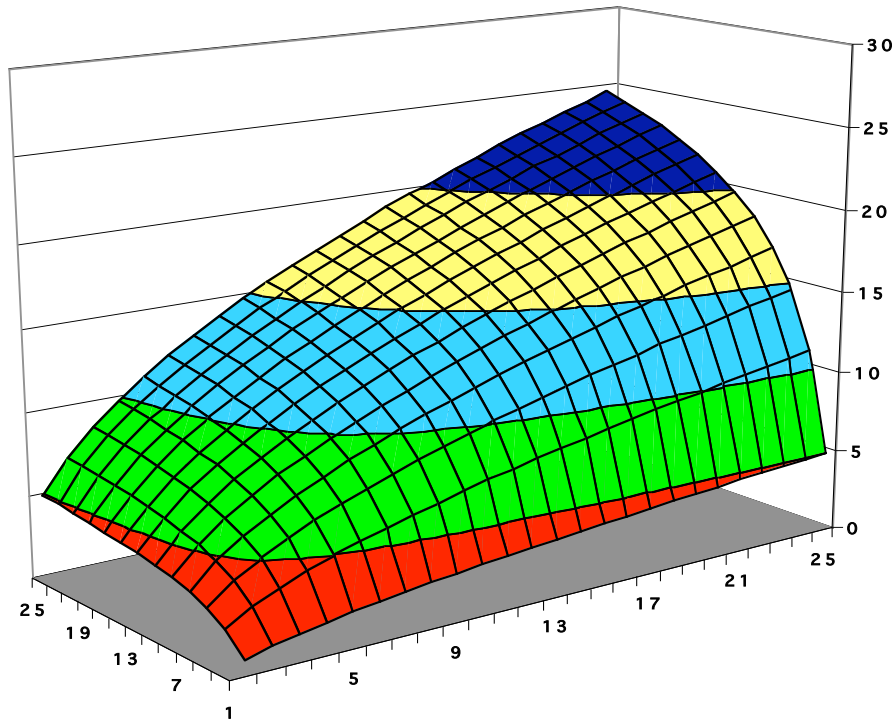
k	Q(k)	k	Q(k)	k	Q(k)	k	Q(k)	k	Q(k)
1	0.3989	11	1.5865	21	1.8892	31	2.0565	41	2.1707
2	0.681	12	1.6293	22	1.9097	32	2.0697	42	2.1803
3	0.8881	13	1.668	23	1.9292	33	2.0824	43	2.1897
4	1.0458	14	1.7034	24	1.9477	34	2.0947	44	2.1988
5	1.1697	15	1.7359	25	1.9653	35	2.1066	45	2.2077
6	1.2701	16	1.766	26	1.9822	36	2.1181	46	2.2164
7	1.3534	17	1.7939	27	1.9983	37	2.1293	47	2.2249
8	1.4242	18	1.82	28	2.0137	38	2.1401	48	2.2331
9	1.4853	19	1.8445	29	2.0285	39	2.1506	49	2.2412
10	1.5389	20	1.8675	30	2.0428	40	2.1608	50	2.2491

Equation 5 applies to systems wherein the modules are asymmetric. If modules are symmetric, then it will be optimal to run the same number of experiments on each module. The 2J arguments in equation 5 then collapse to two, and the value of the system as a whole, denoted $V(j,k)$, becomes:

$$V(j,k) = \sqrt{N_j} Q(k) \quad (6)$$

Figure 10 graphs this function for different values of j and k . The way to read this chart is as follows. The vertical axis shows system value as a function of two variables. The first variable, on the right-hand axis, is the number of modules in the system. In the figure, this variable ranges from 1 to 25. The second variable, on the left axis, is the number of design experiments, i.e. R&D projects, per module. This variable also ranges from 1 to 25. The surface shown on the vertical axis indicates the value of different combinations of modules and experiments. As we go out along the middle of this surface, we see the value of running one experiment on one module, two experiments on each of two modules, and so on, until at the far corner, we have 25 experiments on each of 25 individual modules in the system.

Figure 10
The Value of Splitting and Substitution



The figure shows that there is strong complementarity between design modularity and experimentation. More modules make more experiments more valuable, and vice versa. The two things go together.

Although this function is the result of a thought experiment (and no real system is symmetrical to this degree), the result is nevertheless compelling. The amount of economic value being created by the combination of modules and experiments is really very large. The values are calculated relative to the value of a single experiment in a non-modular system; we see, in effect, that a complex project organized as 25 modules with 25 experiments per module can obtain approximately 25 times the value of the same project organized as a single, interdependent whole. These values are hypothetical, and do not recognize the costs of creating a modular architecture, running experiments, or integrating and testing the modular system. But a value multiple of 25 will pay for a lot of engineering costs! In other words, the incentives

afforded by the combination of modularity and experimentation are so great that in a free-market economy, if it is possible to modularize, someone will do so in order to capture this value.

The graph is also surprising in the way the factors interact. Modules and experiments work together, and therefore, as one increases modularity, and one should experimentation, that is, R&D, as well, in order to get the best outcome. This is costly, as we shall see. But there is another kind of “cost” involved: the cost of potential innovation ignored. Often engineers with the encouragement of managers will modularize a complex engineering system in order to reduce its cognitive complexity or shorten its development time. Neither group may understand what the modularization implies for the value of innovation and experimentation within the system. Yet if the firm that initially modularizes the system ignores this value, *there will be strong incentives for other firms to enter the market offering their own new module designs.*

Something like that happened to IBM after the company introduced the first modular computer, System/360, in the mid-1960s. System/360 was a powerful and popular modular system, a *tour de force* in terms of product design, marketing, and manufacturing. However, IBM’s top managers did not understand the value of the options that had been created by its own new modular design. They did not increase inhouse design efforts, and as a result, left profitable opportunities “on the table.” Before long, new firms moved in and seized these opportunities. This was the start of the pattern of industry evolution and value migration depicted at the beginning of this paper.

Many of the new firms that entered the computer industry in the wake of System/360 were staffed by former IBM engineers. Engineers who had worked on System/360 and its successors could see the module options very well, and knew the design rules of the system. Thus when IBM’s top managers did not fund their projects, they took those projects elsewhere. Beginning in the early 1970s, scores and then hundreds of engineers left IBM and joined others in founding companies that supplied “plug-compatible” modules for IBM’s System/360 and 370. As it happens, one of IBM’s main R&D labs was located in San Jose, California, and the exodus of engineers from the San Jose labs was one of the key factors that contributed to the emergence of what we now call Silicon Valley.

The Costs of Modularity

We have yet to address the costs of modularity. In fact, the costs of creating and exploiting a modular system can be a significant offset to the value that is created.

There are, first of all, the costs of making an interdependent system modular: the cost of creating and disseminating design rules. The DSM mapping techniques, discussed above, show how painstaking the process of modularization must be, if it is to succeed in creating truly independent modules. *Every important cross-module dependency must be understood and addressed via a design rule.* Obviously the density of the dependencies matters here. Modular breakpoints (interfaces) need to go at the “thin crossing points” of the interdependency graph. Some systems are naturally more “loosely-coupled” than others: they have more thin crossing points hence relatively more potential modules.

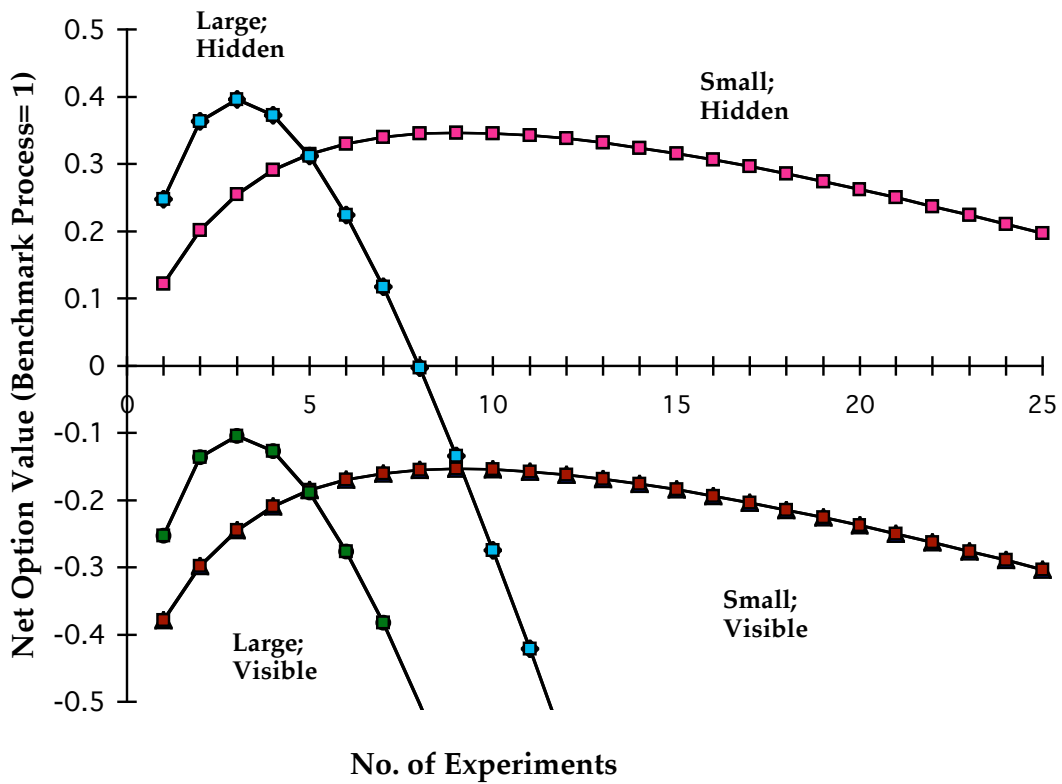
For example, circuits, the physical system on which computers are based, are one- and two-dimensional; whereas mechanical solids are three-dimensional. Clearly it is harder to split up complex, curved, 3-dimensional designs, and to create flexible interfaces for them: there are more dependencies to manage, and the tolerances are much tighter. Thus modularizing an automobile’s design is a tougher engineering problem than modularizing a circuit design: the cost of creating a modular architecture and related interfaces will be higher. This has led some scholars, like Daniel Whitney at MIT, to predict that autos and airplanes will achieve only limited modularity in practice.¹⁹ The option values inherent in these tightly coupled systems will be low relative to systems that can more easily be modularized.

It is also costly to run the experiments needed to realize the potential value of a modular system and to design the tests needed to determine if particular modules are compatible with the system, and which one performs best. Indeed figuring out how many and what kind of experiments to run and how to test the results are important sub-problems within the overall option valuation problem. The interaction of option value and the costs of experimentation and testing modules causes each module in a large system to have a unique value profile. For example, Figure 11 shows how value profiles may differ across modules as a function of the number of experiments. Some modules (those that are hidden and have high technical potential) can support a lot of experiments; others (those that are visible and/or have low technical potential) can support few or none.

¹⁹ Whitney, 1996.

Figure 11
The Value Profiles for Different Modules of a Computer System

Size	Visibility	Examples
large	hidden	disk drive; large application program
large	visible	microprocessor; operating system
small	hidden	cache memory; small application program
small	visible	instruction set; internal bus



Conclusion

We conclude this brief excursion into the realm of modularity by returning to the question of industry evolution, and commenting on the perils of modularity for incumbent firms. Modularity in the design of a complex engineering system with high technical potential (high \square s in the modules) is likely to be highly disruptive to the pre-existing industry structure. Modularity-in-design allows users or system integrators to mix and match of the best designs within each module category and to incorporate new and improved module designs as they become available. Thus a modular system design requires that a company operate all aspects of its business more efficiently than its competitors. If it is not “the best” in a given module, then competitors will flock to that point of vulnerability.

For example, consider IBM’s introduction of the personal computer (PC) in the early 1980s.²⁰ By this time, IBM had learned the basic lessons of modularity inherent in System/360. Its managers understood how modularity encouraged both innovation and entry on modules. Indeed, the PC was extremely modular-in-design, and IBM leveraged this modularity by outsourcing most of its hardware and software components. But IBM’s managers also understood that they needed to protect the company’s privileged position within the modular architecture. Thus IBM retained control of what were thought to be the key design and process modules—a chip called the BIOS (Basic Input Output System) and the manufacturing process of the PC itself. By exercising control of these critical and essential “architectural modules,” IBM’s managers believed that they could manage the rate of innovation in PC’s. Their goal was to obtain maximum return from each generation of PC, before going on to the next.

However, the founders of Compaq had a different idea. First, they independently and legally replicated IBM’s technical control element, the BIOS. Then they designed a machine that was fully compatible with IBM’s published and non-proprietary specifications. They bought the key modules of the PC — the chip and the operating system — and all the other parts they needed from IBM’s own suppliers. And they went to market with an IBM-compatible PC built around the newer, faster Intel 386 chip while IBM was still marketing 286 machines. Within a year Compaq had sales of \$100 million; by 1990, its revenues were \$3 billion and climbing, and IBM was looking to exit from its unprofitable PC business.

²⁰ This history is recounted in Ferguson and Morris, 1994.

However, as everyone knows, the leading player in the PC market today is not Compaq but Dell.²¹ In a nutshell, Dell did to Compaq what Compaq did to IBM: it took advantage of the benefits of modularity and designed a line of technologically competitive, yet compatible, and lower-priced PCs. In fact, Dell used *modularity-in-the-design-of-production-processes* more effectively than Compaq in order to arrive at a more efficient, less-asset-intensive business model. Dell outsourced even more of its manufacturing activities than Compaq. It designed its assembly processes to build machines to order (BTO), thereby cutting out most of its inventory. And it sold its products directly to consumers, thereby cutting out dealers' margins and more inventory. Compaq simply could not compete. When its top managers saw the handwriting on the wall in PCs in early 1998, they tried to move the company into higher-margin "enterprise computing" through the acquisition of Digital Equipment Corporation, but the transition was not successful. Today, Compaq no longer exists as a separate company—it was acquired by Hewlett Packard Corporation in 2002.

In conclusion, the widespread adoption of modularity-in-design in complex engineering systems with high technical potential can set in motion an uncontrollable process of design and industry evolution. The economic consequences of this process—for good and bad—are depicted in the charts and turnover rates described at the beginning of this paper. Thus modularity-in-design can open the door to an exciting, innovative, but very Darwinian world in which no one really knows which firms or business models will ultimately prevail. This can be a world of growth, innovation and opportunity. But, as the Internet Bubble and Crash taught us, this world can also fall into periods of extreme value destruction, chaos and inefficiency. Thus in the last analysis, modularity-in-design is neither good or bad. Rather, it is potentially powerful and disruptive, and therefore dangerous to ignore.

²¹ For detailed comparisons of Compaq's and Dell's business models, see Wheelwright and Verlinden, 1998 and Baldwin and Feinberg, 1999.

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