System Dynamic Model for Computer Virus Prevalance

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ABSTRACT

Regardless of how sophisticated anti-virus technology may become, computer viruses will forever remain in an uneasy coexistence with us and our computers. Individual strains will wax and wane, but as a whole, computer viruses and anti-virus technology will co-evolve much as biological parasites and hosts do. "Dynamics of Computer Virus Prevalence" remains a challenging research area for modeling and simulation, in an attempt to devise policies, and guidelines to control virus prevalence and spreading in organizations.

Standard SIR Epidemic models is used as the backbone of this study, capitalizing on the dual nature of the biological Virus, and computer Virus Prevalence. The basic SIR Model, has been extended to account for an important concept of a "Kill signal" which is generated as an Infected machine is virus-cleared. The "Kill Signal" Spreads through a "Word of Mouth Contact Rate" informing physically connected machines, or machines that has exchanged Software with the infected one about the virus. The Mathematical Model of the extended SIR is formulated with some details.

As a Policy Design it is suggested that the Kill Signal Level, represented by a Word of Mouth Contact Rate (WMCR) to be dynamically increasing as the Infectious Population level Increases. This has proven theoretically using the Simulation model to push down the Epidemic Level significantly, and thus suggesting a very cost effective organizational policy to control the computer Virus Spreading. On the prevention level policy, the authors have suggested, controlling the infectivity as well as the contact rate (through network topology or transfer media) will help decrease the infection rate of the PC's.

Keywords: Computer Viruses, Epidemiological, SIR Model

1 INTRODUCTION

Computer viruses have pervaded popular culture at least as successfully as they have the world's computer population. Computer viruses have become the subject of widespread urban legends and hoaxes, popular television shows and movies. Yet they have not received much scientific scrutiny.

Much of their popular presence is attributable to an obvious but deep <u>biological analogy</u>: computer viruses replicate by attaching themselves to a host (a program or computer instead of a biological cell) and co-opting the host's resources to make copies of themselves [3].

Biological analogy has been found to be key in understanding the propagation of computer viruses on a global scale and inspirational in the development of defenses against them.

Computer viruses can trace their pedigree to John von Neumann's studies of self-replicating mathematical automata in the 1940s. Although the idea of programs that could infect computers dates to the 1970s, the first well-documented case of a computer virus spreading "in the wild" occurred in 1986 [6], when a code snippet known as the "Brain" virus appeared on several dozen diskettes at the University of Delaware. Today viruses afflict at least a million computers every year. Users spend several hundred million dollars annually on antivirus products and services, and this figure is growing rapidly.

Appendix B, demonstrates a very interesting Time-Line History for Virus Prevalence starting from 1949 up to date.

There are three main classes of PC viruses, file infectors, boot-sector viruses and macro viruses.

File Infectors: Roughly 85 percent of all known viruses infect files containing applications such as spreadsheet programs or games. When a user runs an infected application, the virus code executes first and installs itself independently in the computer's memory so that it can copy itself into subsequent applications that the user runs. Once in place, the virus returns control to the infected application; the user remains unaware of its existence. Eventually a tainted program will make its way to another computer via a shared diskette or network, and the infection cycle will begin anew.

Boot-sector viruses: Accounts for about 5 percent of known PC virus strains, reside in a special part of a diskette or hard disk that is read into memory and executed when a computer first starts. Once loaded, a boot-sector virus can infect any diskette that is placed in the drive. It also infects the hard disk, so that the virus will be loaded into memory whenever the system is restarted. Boot viruses are highly effective: even though there are fewer strains, they were for a time much more prevalent than file infectors were.

Macro viruses: The third category, macro viruses, are independent of operating systems and infect files that are usually regarded as data rather than as programs. Many spreadsheet, database and word-processing programs can execute scripts--prescribed sequences of actions-embedded in a document. Such scripts, or macros, are used to automate actions ranging from typing long words to carrying out complicated sequences of calculations.

Antivirus Technology: Antivirus software has existed since shortly after computer viruses first appeared. Generic virus-detection programs can monitor a computer system for virus-like behavior (such as modification of certain crucial files or parts of main memory), and they can periodically check programs for suspicious modifications. Such software can even detect hitherto unknown viruses, but it can also be prone to false alarms because some legitimate activities resemble viruses at work.

2 PROBLEM DEFINITION AND OBJECTIVE THE PAPER

With the resurgence of the Internet, computer viruses are able to propagate much faster, and more aggressively. As early as 1988, Robert Tappan Morris launched what came to be known as the "Internet Worm," a program that exploited security holes and invaded hundreds of computers around the world in less than a day. Also New type of viruses Like "Melissa", and Love Bug Spread through Outlook Contacts by replicating Fake e-mails.

Most Researches on Computer Viruses [4] are on the Microscopic Level (Virus structure, Replication mechanism, Type, and extent of damage, ... etc).

The focus of this study, however, is more on the macroscopic level, where Dynamics models for Computer Virus Prevalence inspired from Mathematical epidemiological models are presented.

Standard Epidemiological SIR Model is being presented, however a logical and practical extension has been introduced for the case of computer viruses. This extension accounts for a "Word of Mouth Contact Rate WMCR" or also known as "Kill Signal", whereby a known detected Virus is associated by Central reporting and spreading of "Word of Mouth" awareness to other computer users, with direct adequate contact with the Infected Machine.

It will be demonstrated that this Central Reporting and Word of mouth spreading portrays a very important message, and a Policy design criteria, perhaps more important than relying on the latest anti-virus technology, in order to push down the virus below epidemic threshold.

Mathematical formulation for the dynamics of Computer Virus Prevalence will be presented. Reference Modes for Various types of Computer Viruses are introduced, and used for testing the model. Causal Loop Diagram is designed and used to develop, a Stock and Flow model used to simulate the Problem. Various Sensitivity Analyses are explored, and Finally Policy Decisions are discussed and applied to the Simulation model.

3 MATHEMATICAL FORMULATION

3.1 Basic SIR Model

In our modeling of computer virus spread [1] some important concepts and simplifications have been borrowed from the well-established field of mathematical epidemiology [3].

In particular, the standard SIR (Susceptible, Infectious, and Recovered) Model [1] is used as the base for the mathematical Formulation.

Standard SIR model is shown as follows in Fig. 1

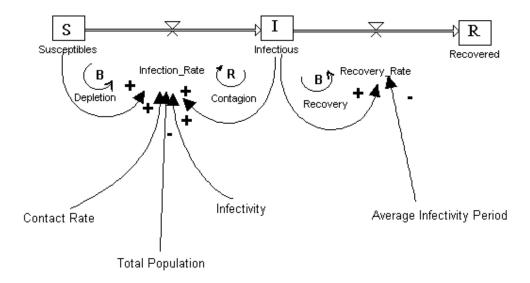


Figure 1: Standard SIR Epidemiological Model

Starting from a Total population of N objects (in this case PCs), Contact Rate "c" is defined and measured as Computers Interfaced or contacted per unit of Computer Population per time.

Contact Rate "c" is mainly a function of Network topology and Nature of Computers Interconnections. The term "contact" here, does not only imply physical contact but also through Software Exchange (for e.g. Diskettes exchange). In that case contact rate implies rate of S/W workers exchanging Software per population machine per time.

Infectivity i is a dimensionless Factor that represents the probability that a susceptible machine gets infected when connected to an infectious machine (or to the network).

In the SIR model it is assumed that the Infected machine remains infected for a certain period, which is the Average Infectivity period " d_i ", after which the machine is scanned for the virus, and then cleared. The Model does not take into consideration that the machine can be reformatted, this is considered as out of the scope of the research.

The parameter d_i affects the flow rate from the Infectious population of PC's to the Recovered Population.

With Reference to the SIR Model, the Susceptibles, Infectious, and Recovered PC Population are governed by the Relationship;

$$S + I + R = N \tag{1}$$

The dynamics of the above SIR model is governed by the set of Non-Linear coupled Differential Equations described as follows,

$$\frac{dI}{dt} = (IR - RR) \tag{2}$$

$$\frac{dR}{dt} = -RR\tag{3}$$

where IR, and RR are the Infection Rate, and Recovery Rate, respectively.

$$IR = \frac{ciS}{N} \tag{4}$$

$$RR = \frac{I}{d_i} \tag{5}$$

Substituting Equations 4, and 5 into equations 2, and 3, and substituting for *S* in Equation 4 for *I*, and *R* in Equation 1, we get,

$$\frac{dI}{dt} = \frac{ciI(N - I - R)}{N} - \frac{I}{d_i} \tag{6}$$

and,

$$\frac{dR}{dt} = \frac{I}{d_i} \tag{7}$$

3.2 Extended SIR Model

The basic SIR Model represented above has been extended [3,5], to account for a practical observation that exists in the SW Workers Community, which will lead to the concept of Word of Mouth Contact Rate (WMCR), or kill signal discussed earlier.

This is the concept of a Word of Mouth Contact Rate (WMCR), that starts spreading much like an epidemic. When a computer virus is detected, SW worker starts informing other Workers who have exchanged SW in the last period, and contacts them verbally or by e-mails to warn them for the possibility of a certain virus prevalence. This as will be depicted later in the simulation will have a great impact in reducing the original virus Spreading, and will thus represent a key Policy Decision Criteria for pushing down the virus below the epidemic threshold.

Mathematically, the WMCR will be represented by " c_k ", where the Subscript k represents a "kill" signal. Hence WMCR, c_k will be defined as the number of colleagues contacted (assuming each colleague has one machine), per Machines Population per time.

The Introduction of c_k lowers the Infection Rate by a Quantity which is proportional to the Current Infectious, and Recovered Population, given by, $c_k IR/N$

Accordingly Equation (6) will be modified to,

$$\frac{dI}{dt} = \frac{ciI(N - I - R)}{N} - \frac{I}{d_i} - \frac{c_k IR}{N}$$
(8)

In the extended SIR model we do not assume that the recovered machine will remain for Eternity, but after a certain period of time will be susceptible to new viruses again. This is the Average Recovery period " d_r ", after which the machine is susceptible to new viruses. This parameter d_r affects the flow rate from the Recovered population to the Susceptible Population. The effect of d_r is accounted for in Equation (7) as follows,

$$\frac{dR}{dt} = \frac{c_k IR}{N} - \frac{R}{d_r} + \frac{I}{d_i} \tag{9}$$

3.3 Endogenous Focus and Causal Loop Analysis

In what follows the dynamic hypothesis is explained in terms of the endogenous consequences of Feedback Structures.

As shown in the Accompanying Causal Loop Diagram, 6 Causal loops are inherent in our model. Three loops are positive Re-enforcing Loops, and three are negative Balancing. These Loops interact, to generate the dynamic Behavior that simulates the reference modes as will be presented in simulation and Testing section.

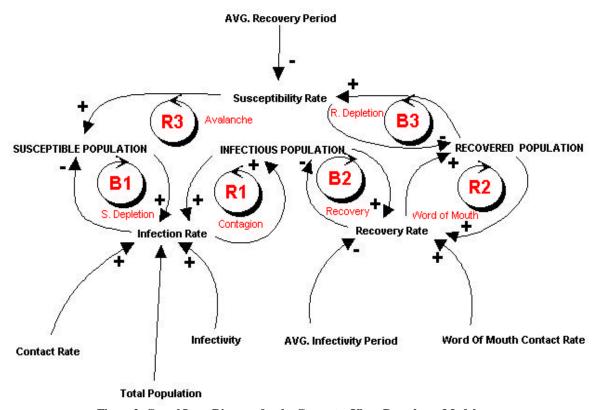


Figure 2: Causal Loop Diagram for the Computer Virus Prevalence Model

In what follows each loop is analyzed,

B1: S. Depletion Loop

In this loop, S. Stands for Susceptible Population, accordingly this loop main function is to deplete the Susceptible Population Stock. As the Susceptible Population Increase the Infection Rate Increases, and as the Infection Rate Increases (outflow with respect to the susceptible population), the Susceptible Population is depleted.

R1: Contagion Loop

In this loop, this loop main function is to raise the Infection Population. As the Infection Population Increases, the Infection rate Increases, which increases more the Infection Population.

B2: Recovery Loop

The main function of this loop is to deplete the Infectious Population Stock, and accordingly increase the Recovered Population. As the Infectious Population Increase the recovery Rate Increases, and as the recovery Rate Increases (outflow with respect to the infectious population), the Infectious Population is depleted, and the recovered Population rises.

R2: Word of Mouth Loop

The main function of this loop is to raise the Recovered Stock. This is the key contribution in this study, is that the Word of Mouth Spreading that a Virus has been detected by machine speeds up Recovery, since other contacted/connected machine will be aware, and start detecting and Clearing the Virus. As the recovered population increases, the recovery Rate Increases through the Word Mouth Effect, and as the recovery Rate Increases, then the Recovered Population increases.

B3: R. Depletion Loop

In this loop, where R. Stands for Recovered Population, accordingly this loop main function is to deplete the Recovered Population Stock. As the level of the Recovered Population Increases, the Susceptibility Rate Increases, and as the Susceptibility Rate Increases (outflow with respect to the recovered population), the Recovered Population is depleted.

R3: Avalanche Loop

This Loop tends to re-enforce all populations. In case of an open Community, and assuming that this loop is dominant, it tends to increase exponentially, and indefinitely the levels of all population stocks, namely the Susceptible, Infectious, and Recovered. In our model however, we assume a closed community, accordingly, the Total Sum of population is N, as in Equation (1). This Loop actually passes through all endogenous Variables.

The Following is the corresponding Stock and Flow Simulation diagram Figure 3. Appendix B shows the stock and flow equations used in the model.

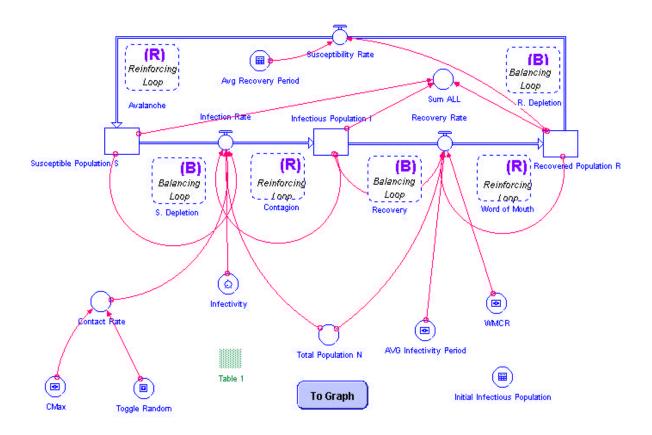


Figure 3: Stock and Flow Diagram for the Computer Virus Prevalence Model

4 MODEL SIMULATION RESULTS

The Following Figure shows a Typical Run for the Infection Rate.

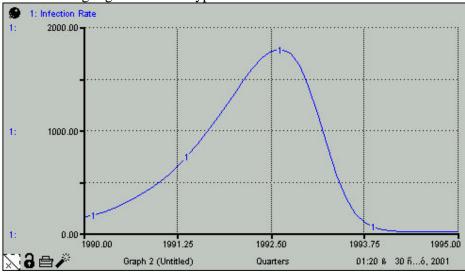


Figure 4: Infection Rate (I)

The Infection Rate as shown in Figure 4 exhibits an exponential rise followed by a decline due to the interaction of several loops, as discussed in the Dynamic Hypothesis In the above Run the Following data were used,

• Total Population N: 1000

Infectivity: 30%Contact Rate: 5WMCR: 16

Average Infectivity Period: 8Average Recovery Period: 10

In the full paper, sensitivity analysis is being performed and demonstrated using key model parameters

5 POLICY DESIGN AND EVALUATION

Word of Mouth Contact plays a significant role in pushing down, and slowing down the Viral Epidemic. Several Runs of the simulation model for various values of WMCR while observing the Infectious population peak, has revealed a rather negative exponential relationship, as shown in the next Figure 5.

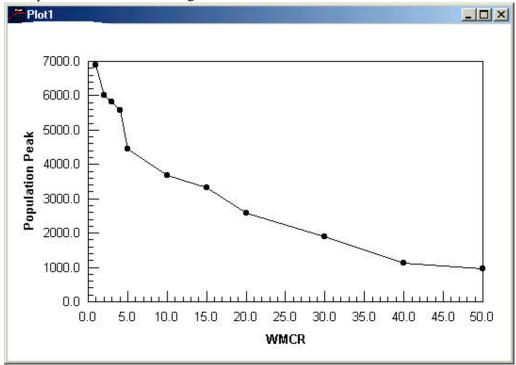


Figure 5: Population Peak as a Function of WMCR

This relationship fits to a good extent the following Formula,

Infectious Population =
$$6593 \exp(-0.046819*WMCR)$$
 (10)

Therefore, a suggested Policy is to derive a negative Feedback loop through a positive relationship between the Level of Infectious Population Stock and WMCR. Hence WMCR will increase as the Infectious population Increases.

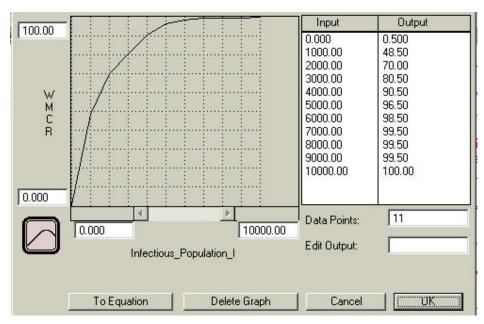


Figure 6: Policy Design Functional Relationship

Applying this suggested Policy on the Simulation model, shows a remarkable effect on the Epidemic by pushing down the peak value, and decreasing the epidemic Duration as shown in next figure.

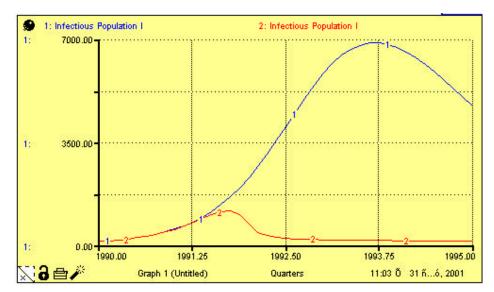


Figure 7: Infectious Population after Applying Policy Decision

Run #1 Infectious Population without the application of the suggested Policy Run #2 Infectious Population after applying the Suggested Policy.

The above figure shows the importance of the Word of Mouth Contact rate, and the concept of the Kill signal to push down the Computer Viral Epidemic to a great extent. This suggests that central reporting of virus incidence, followed by dissemination of these findings to largest possible number of related machines, in addition to the users awareness, and responsiveness will alleviate to a great extent the computer virus Epidemic Effect.

On the prevention level policy, the authors have suggested, controlling the infectivity as well as the contact rate (through network topology or transfer media) will help decrease the infection rate of the PC's. However, this suggestion has been tested and attempted.

6 SUMMARY AND CONCLUSION

Background Information is overviewed covering various types of viruses, and anti-virus technologies. An interesting time line of most common viruses is shown in Appendix A.

Mathematical formulation of the problem is presented with some details. Standard SIR Epidemic model was used as the backbone of the study. The main contribution in this paper is developing and simulating an extended SIR model using STELLA Software. The concept of "Kill Signal" represented by a Word of Mouth Contact Rate Parameter is introduced. The basic idea is that as much as the computer virus starts exercising it's epidemic behavior, and as computer viruses are detected, and cleared, Kill signal starts spreading through a word of mouth to machines that has been contacted or exchanged Software with the infected machine. The governing equations that dictate this effect have been also incorporated in the model.

The Standard Steps in the modeling process are followed. Starting from Problem articulation to Dynamic hypothesis to testing and simulation.

Time Horizon is selected to be in the period 1990-1995, in order to test and verify against the reference modes that was collected from the Literature in this period. (This is shown in the full paper)

Causal Loop Diagram of the problem, and the associated endogenous focus, has been presented and explained. A Complete workable Stock and flow model has been built using Stella Software.

The observed behavior of the Infection rate for all viruses showed, an epidemic rise followed by a decline, which has been verified in the Simulation Model.

Sensitivity Analysis of Key exogenous variables has shown logical consistency as for the Expected sensitivity results.

The main focus in the Sensitivity Analysis was on the Effect of WMCR (Kill Signal) in slowing down the extent of the Epidemic. It has been demonstrated that as the WMCR increases the peak of the Epidemic decreases exponentially by data fitting this relationship.

As a Policy Design a feedback mechanism (another Balancing loop) is suggested, where the WMCR is dynamic and increases as the Infectious Population level Increases. The dynamic reaction of the WMCR means as the Infectious population increases the extent of Virus Central Reporting should widen to inform more users about Virus Existence. This slowed down significantly the virus epidemic level.

These theoretical results on kill signals (WMCR) are exciting because they suggest a very cost-effective technique for thwarting viral spread. A number of different implementations can be considered, including user education (getting people to tell their friends if they discover a computer virus) and organizational policies which encourage users to report virus

incidents to a central agency, which can then ensure that machines in the vicinity of the infected machine are scanned for viruses and cleaned up.

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APPENDIX A: STOCK AND FLOW EQUATIONS

 $\label{local_population_I(t) = Infectious_Population_I(t - dt) + (Infection_Rate - Recovery_Rate) * dt INIT Infectious_Population_I = Initial_Infectious_Population$

INFLOWS

Infection Rate=

Contact_Rate*Susceptible_Population_S*Infectious_Population_I*Infectivity/Total_Population_N

OUTFLOWS

Recovery_Rate=

Infectious_Population_I/AVG_Infectivity_Period+Infectious_Population_I*Recovered_Population_R*WMCR/Total Population N

 $Recovered_Population_R(t) = Recovered_Population_R(t - dt) + (Recovery_Rate - Susceptibility_Rate) * dt INIT Recovered_Population_R = 0$

INFLOWS

Recovery_Rate=

Infectious_Population_I/AVG_Infectivity_Period+Infectious_Population_I*Recovered_Population_R*WMCR/Total_Population_N

OUTFLOWS

Susceptibility_Rate = Recovered_Population_R/Avg_Recovery_Period

 $Susceptible_Population_S(t) = Susceptible_Population_S(t-dt) + (Susceptibility_Rate-Infection_Rate) * dt INIT Susceptible_Population_S = Total_Population_N-Infectious_Population_I-Recovered_Population_R$

INFLOWS

 $Susceptibility_Rate = Recovered_Population_R/Avg_Recovery_Period$

OUTFLOWS

Infection_Rate=

Contact Rate*Susceptible Population S*Infectious Population I*Infectivity/Total Population N

 $AVG_Infectivity_Period = 2$

Avg_Recovery_Period = 2

CMax = 7

Contact_Rate = IF(Toggle_Random=1) THEN RANDOM(1,CMax) ELSE CMax

Epidemic_Threshold = Infection_Rate/Recovery_Rate

Infectivity = 1

Initial_Infectious_Population = 0.01

Sum ALL = Infectious Population I+Recovered Population R+Susceptible Population S

 $Toggle_Random = 1$

Total Population N = 10000

WMCR = 0.5

APPENDIX B: TIME LINE OF COMPUTER VIRUSES (1949 – TO DATE)

1949	Theories for self-replicating programs are first developed.
1981	Apple Viruses 1, 2, and 3 are some of the first viruses "in the wild," or public domain. Found on the Apple II operating system, the viruses spread through Texas A&M via pirated computer games.
1983	Fred Cohen, while working on his dissertation, formally defines a computer <u>virus</u> as "a computer program that can affect other computer programs by modifying them in such a way as to include a (possibly evolved) copy of itself."
1986	Two programmers named Basit and Amjad replace the executable code in the boot sector of a <u>floppy disk</u> with their own code designed to infect each 360kb floppy accessed on any drive. Infected floppies had "© Brain" for a volume label.
1988	One of the most common viruses, Jerusalem, is unleashed. Activated every Friday the 13th, the virus affects both .EXE and .COM files and deletes any programs run on that day.
1990	Symantec launches <u>Norton AntiVirus</u> , one of the first anti-virus programs developed by a large company.
1991	Tequila is the first widespread <u>polymorphic</u> virus found in the wild. Polymorphic viruses make detection difficult for virus scanners by changing their appearance with each new infection.
1992	1300 viruses are in existence, an increase of 420% from December of 1990. The Michelangelo scare predicts 5 million computers will crash on March 6. Only 5,000–10,000 actually go down.
1994	Good Times <u>email hoax</u> tears through the computer community. The hoax warns of a malicious virus that will erase an entire hard drive just by opening an email with the subject line "Good Times." Though disproved, the <u>hoax</u> resurfaces every six to twelve months.
1998	Currently harmless and yet to be found in the wild, StrangeBrew is the first virus to infect Java files. The virus modifies CLASS files to contain a copy of itself within the middle of the file's code and to begin execution from the virus section.

1999

The Melissa virus, W97M/Melissa, executes a macro in a document attached to an email, which forwards the document to 50 people in the user's Outlook address book. The virus also infects other Word documents and subsequently mails them out as attachments. Melissa spread faster than any other previous virus.

The Love Bug, also known as the ILOVEYOU virus, sends itself out via Outlook, much like Melissa. The virus comes as a VBS attachment and deletes files, including MP3, MP2, and JPG. It also sends usernames and passwords to the virus' author.

2000

W97M.Resume.A, a new variation of the Melissa virus, is determined to be in the wild. The "resume" virus acts much like Melissa, using a Word macro to infect Outlook and spread itself.

The "Stages" virus, disguised as a joke email about the stages of life, spreads across the Internet. Unlike previous viruses, Stages is hidden in an attachment with a false ".txt" extension, making it easier to lure recipients into opening it. Until now, it has generally been safe to assume the text files are safe.

The Anna Kournikova virus, also known as VBS/SST, which masquerades as a picture of Tennis Star Anna Kournikova, operates in a similar manner to Melissa and The Love Bug. It spreads by sending copies of itself to the entire address book in Microsoft Outlook. It is believed that this virus was created with a so-called virus creation kit, a program which can enable even a novice programmer to create these malicious programs.

2001

In May, the HomePage email virus hit no more than 10,000 users of Microsoft Outlook. When opened, the virus redirected users to sexually explicit Web pages. Technically known as VBSWG.X, the virus spread quickly through Asia and Europe, but was mostly prevented in the U.S. because of lessons learned in earlier time zones. The author of the virus is said to live in Argentina, and have authored the Kournikova virus earlier in the year.

The Code Red I and II worms attacked computer networks in July and August. According to Computer Economics they affected over 700,000 computers and caused upwards of 2 billion in damages. A worm spreads through external and (then) internal computer networks, as opposed to a virus which infects computers via email and certain websites. Code Red took advantage of a vulnerability in Microsoft's Windows 2000 and Windows NT server software. Microsoft developed a patch to protect networks against the worm, and admits that they too were attacked. Other major companies affected include AT&T, and the AP.