Analysis and	Mitigation	of Multiple	Radiation	Induced	Errors in	Modern C	Circuits

by

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M.Sc., Electrical & Computer Engineering, Southern Illinois University, 2012

A Dissertation Submitted in Partial Fulfillment of the Requirements for the Master of Science Degree

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DISSERTATION APPROVAL

TITLE (ANALYSIS AND MITIGATION OF MULTIPLE RADIATION INDUCED ERRORS IN MODERN CIRCUITS)

By

Adam Watkins

A Dissertation Submitted in Partial

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Approved by:

Dr. Spyros Tragoudas, Chair

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Graduate School Southern Illinois University Carbondale (Date of Approval) AN ABSTRACT OF THE DISSERTATION OF

Adam Watkins, for the Doctor of Philosophy degree in Electrical and Computer

Engineering, presented on November 2nd, at Southern Illinois University Carbondale. (Do

not use abbreviations.)

TITLE: ANALYSIS AND MITIGATION OF MULTIPLE RADIATION INDUCED ER-

RORS IN MODERN CIRCUITS

MAJOR PROFESSOR: Dr. S. Tragoudas

Due to technology scaling, the probability of a high energy radiation particle striking

multiple transistors has continued to increase. This, in turn has created a need for new

circuit designs that are can tolerate multiple simultaneous errors. A common type of error

in memory elements is the double node upset (DNU) which has continued to become more

common. All existing DNU tolerant designs either suffer from high area and performance

overhead, may lose the data stored in the element or are vulnerable to an error after a DNU

occurs which makes the devices unsuitable for clock gating. In this dissertation, a novel

latch design is proposed in which all nodes are capable for fully recovering their correct

value after a single or double node upset which is referred to as DNU robust. The proposed

latch offers lower delay, power consumption and area requirements compared to existing

DNU robust designs.

Multiple simultaneous radiation induced errors are a current problem that must be

studied in combinational logic. Typically, simulators are used early in the design phase

which use a netlist and rudimentary information of the process parameters to determine

the error rate of a circuit. Existing simulators are able to accurately determine the effects

when the problem space is limited to one simultaneous error. However, existing methods do

not provide accurate information when multiple concurrent errors occur due to inaccurate

ii

approximation of the glitch shape when multiple errors meet at a gate. To improve existing error simulation, a novel analytical methodology to accurate determine the pulse shape when two simultaneous errors occur is proposed. Through extensive simulations, it was shown that the proposed methodology matches closely with HSPICE while providing a speedup of 15X.

The analysis of the soft error rate of a circuit has continued to be a difficult problem due to the calculation of the logical effect on a pulse generated by a radiation particle. The most common existing methods to determine logical effects use either exhaustive input pattern simulation or binary decision diagrams. The problem with both approaches is that simulation of the circuit can intractably time consuming. To solve this issue, a simulation tool is proposed which employs an adaptive partitioning algorithm to reduce the simulation time and space overheads of binary decision diagram based simulation. Compared to existing simulation tools, the proposed tool can simulate larger circuits faster.

DEDICATION

(NO REQUIRED FOR RESEARCH PAPER)

(The dedication, as the name suggests is a personal dedication of one's work. The section is OPTIONAL and should be double-spaced if included in the thesis/dissertation.)

ACKNOWLEDGMENTS

(NOT REQUIRED IN RESEARCH PAPER)

I would like to thank Dr. Jones for his invaluable assistance and insights leading to the writing of this paper. My sincere thanks also goes to the seventeen members of my graduate committee for their patience and understanding during the nine years of effort that went into the production of this paper.

A special thanks also to Howard Anton [1], from whose book many of the examples used in this sample research paper have been quoted. Another special thanks to Prof. Ronald Grimmer who provided the previous thesis template upon which much of this is based and for help with graphics packages.

TABLE OF CONTENTS

A۱	ostrac	t	ii
De	edicat	ion	iv
A	cknow	vledgments	v
Li	st of	Tables	vii
Li	st of 1	Figures	viii
In	trodu	ction	1
1	Syst	tems of Linear Equations and Matrices	2
	1.1	Introductions to Systems of Linear Equations	2
	1.2	Gaussian Elimination	4
	1.3	Further Results on Systems of Equations	5
		1.3.1 Some Important Theorems	5
2	Syst	sems of Linear Equations and Matrices	7
	2.1	Introductions to Systems of Linear Equations	7
	2.2	Gaussian Elimination	9
	2.3	Further Results on Systems of Equations	10
		2.3.1 Some Important Theorems	10
3	Exa	mples	12
Re	eferen	ces	17
Vi	ta.		18

LIST OF TABLES

LIST OF FIGURES

1	A energetic particle creating electron-hole pairs in a transistor	1
1.1	(a) no solution, (b) one solution, (c) infinitely many solutions	3
1.2	Inside and outside numbers of a matrix multiplication problem of $A \times B =$	
	AB, showing how the inside dimensions are dropped and the dimensions of the	
	product are the outside dimenions	5
2.1	(a) no solution, (b) one solution, (c) infinitely many solutions $\dots \dots$	8
2.2	Inside and outside numbers of a matrix multiplication problem of $A \times B =$	
	AB, showing how the inside dimensions are dropped and the dimensions of the	
	product are the outside dimenions	10
3.1	Two rows of graphics: (a) Square (b) Circle (c) Rectangle	13
3.2	Three rows of graphics: (a)–(c) Squares. (d)–(f) Circles. (g)–(i) Ovals	14
3.3	Use of verbatim environment	15
3.4	Matrix Rotated 90 degrees.	16

INTRODUCTION

The continued reduction of the transistor feature size has led to the increase in the vulnerability of modern circuits to error. This effect, in turn, has made modern circuits more susceptible to radiation induced errors in space and terrestrial environments. A radiation induced error, commonly referred to as a soft error, occurs when a high energy particle from space or packaging strikes a transistor. Shown in Fig. 1, the particle deposits energy in the active volume generating electron-hole pairs. This mechanism creates a new diffusion region that could allow a non-conducting device to temporarily conduct current. The mechanism causes a temporary voltage pulse, referred to as a single event transient (SET) to occur at the device.

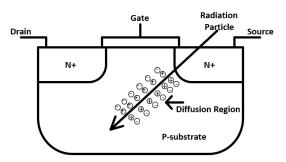


Figure 1. A energetic particle creating electron-hole pairs in a transistor.

In the case of a radiation particle striking a memory storage element, the data stored in the device can be changed. If the device is upset by a single particle, this is referred to as a single event upset (SEU).

CHAPTER 1

SYSTEMS OF LINEAR EQUATIONS AND MATRICES

1.1 INTRODUCTIONS TO SYSTEMS OF LINEAR EQUATIONS

In this section we introduce base terminology and discuss a method for solving systems of linear equations.

A line in the xy-plain can be represented algebraically by an equation of the form

$$a_1x + a_2y = b$$

An equation of this kind is called a linear equation in the variables x and y. More generally, we define a linear equation in the n variables x_1, \ldots, x_n to be one that can be expressed in the form

$$a_1 x_1 + a_2 x_2 + \dots + a_n x_n = b \tag{1.1}$$

where $a_1, a_2, \dots a_n$ and b are real constants.

Definition. A finite set of linear equations in the variables x_1, x_2, \ldots, x_n is called a *system* of linear equations.

Not all systems of linear equations has solutions. A system of equations that has no solution is said to be *inconsistent*. If there is at least one solution, it is called *consistent*. To illustrate the possibilities that can occur in solving systems of linear equations, consider a general system of two linear equations in the unknowns x and y:

$$a_1x + b_1y = c_1$$

$$a_2x + b_2y = c_2$$

The graphs of these equations are lines; call them l_1 and l_2 . Since a point (x, y) lies on a line if and only if the numbers x and y satisfy the equation of the line, the solutions of the

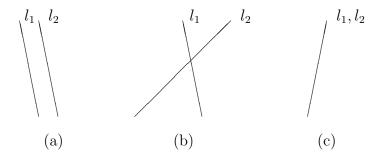


Figure 1.1. (a) no solution, (b) one solution, (c) infinitely many solutions system of equations will correspond to points of intersection of l_1 and l_2 . There are three

possibilities: The three possibilities illustrated in Figure 2.1 are as follows:

- (a) l_1 and l_2 are parallel, in which case there is no intersection, and consequently no solution to the system.
- (b) l_1 and l_2 intersect at only one point, in which case the system has exactly one solution.

(c) l_1 and l_2 coincide, in which case there are infinitely many points of intersection, and consequently infinitely many solutions to the system.

Although we have considered only two equations with two unknowns here, we will show later that this same result holds for arbitrary systems; that is, every system of linear equations has either no solutions, exactly one solution, or infinitely many solutions.

1.2 GAUSSIAN ELIMINATION

In this section we give a systematic procedure for solving systems of linear equations; it is based on the idea of reducing the augmented matrix to a form that is simple enough so that the system of equations can be solved by inspection.

Remark. It is not difficult to see that a matrix in row-echelon form must have zeros below each leading 1. In contrast a matrix in reduced row-echelon form must have zeros above and below each leading 1.

As a direct result of Figure 2.1 on page 8 we have the following important theorem.

Theorem 1.2.1. A homogenous system of linear equations with more unknowns than equations always has infinitely many solutions

The definition of matrix multiplication requires that the number of columns of the first factor A be the same as the number of rows of the second factor B in order to form the product AB. If this condition is not satisfied, the product is undefined. A convenient way to determine whether a product of two matrices is defined is to write down the size of the first factor and, to the right of it, write down the size of the second factor. If, as in Figure 2.2, the inside numbers are the same, then the product is defined. The outside numbers then give the size of the product.

Although the commutative law for multiplication is not valid in matrix arithmetic, many familiar laws of arithmetic are valid for matrices. Some of the most important ones and their names are summarized in the following proposition.

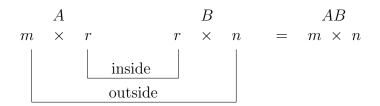


Figure 1.2. Inside and outside numbers of a matrix multiplication problem of $A \times B = AB$, showing how the inside dimensions are dropped and the dimensions of the product are the outside dimensions.

Proposition 1.2.2. Assuming that the sizes of the matrices are such that the indicated operations can be performed, the following rules of matrix arithmetic are valid.

- (a) A + B = B + A (Commutative law for addition)
- (b) A + (B + C) = (A + B) + C (Associative law for addition)
- (c) A(BC) = (AB)C (Associative law for multiplication)

1.3 FURTHER RESULTS ON SYSTEMS OF EQUATIONS

In this section we shall establish more results about systems of linear equations and invertibility of matrices. Our work will lead to a method for solving n equations in n unknowns that is more efficient than Gaussian elimination for certain kinds of problems.

1.3.1 Some Important Theorems

Theorem 1.3.1. If A is an invertible $n \times n$ matrix, then for each $n \times 1$ matrix B, the system of equations AX = B has exactly one solution, namely, $X = A^{-1}B$.

Proof. Since $A(A^{-1}B) = B$, $X = A^{-1}B$ is a solution of AX = B. To show that this is the only solution, we will assume that X_0 is an arbitrary solution, and then show that X_0 must be the solution $A^{-1}B$.

If X_0 is any solution, then $AX_0 = B$. Multiplying both sides by A^{-1} , we obtain $X_0 = A^{-1}B$.

Theorem 1.3.2. Let A be a square matrix.

- (a) If B is a square matrix satisfying BA = I, then $B = A^{-1}$.
- (b) If B is a square matrix satisfying AB = I, then $B = A^{-1}$.

In our later work the following fundamental problem will occur over and over again in various contexts.

Let A be fixed $m \times n$ matrix. Find all $m \times 1$ matrices B such that the system of equations AX = B is consistent.

If A is an invertible matrix, Theorem 2.3.2 completely solves this problem by asserting that for every $m \times n$ matrix B, AX = B has the unique solution $X = A^{-1}B$.

CHAPTER 2

SYSTEMS OF LINEAR EQUATIONS AND MATRICES

2.1 INTRODUCTIONS TO SYSTEMS OF LINEAR EQUATIONS

In this section we introduce base terminology and discuss a method for solving systems of linear equations.

A line in the xy-plain can be represented algebraically by an equation of the form

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An equation of this kind is called a linear equation in the variables x and y. More generally, we define a linear equation in the n variables x_1, \ldots, x_n to be one that can be expressed in the form

$$a_1 x_1 + a_2 x_2 + \dots + a_n x_n = b \tag{2.1}$$

where $a_1, a_2, \dots a_n$ and b are real constants.

Definition. A finite set of linear equations in the variables x_1, x_2, \ldots, x_n is called a *system* of linear equations.

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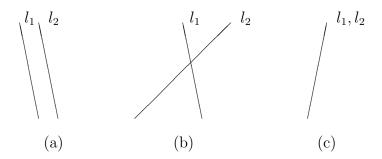


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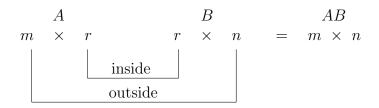


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CHAPTER 3

EXAMPLES

Some examples of the definitions found in the file ps-defs.tex follow below.

Here are examples of how you can use equation numbers with multiple line equations.

$$(f + (g + h))(a) = f(a) + (g + h)(a)$$

$$= f(a) + (g(a) + h(a))$$

$$= (f(a) + g(a)) + h(a)$$

$$= (f + g)(a) + h(a)$$

$$= ((f + g) + h)(a)$$

$$= f(a) + (g + h)(a)$$

$$= f(a) + (g(a) + h(a))$$

$$= (f(a) + g(a)) + h(a)$$

$$= (f + g)(a) + h(a)$$

$$= ((f + g) + h)(a)$$

$$(3.3)$$

$$= (f + g)(a) + h(a)$$

$$= ((f + g) + h)(a)$$

$$= f(a) + (g(a) + h(a))$$

$$= f(a) + (g(a) + h(a))$$

$$= (f(a) + g(a)) + h(a)$$

$$= (f + g)(a) + h(a)$$

$$= ((f + g)(a) + h(a))$$

Below is a figure which shows how to line up small figures on multiple lines. The .dvi version is immediately below. The .pdf version may be found underneath the complete figure and commented out. If you exchange the sections commented out, then you can compile a .pdf file.

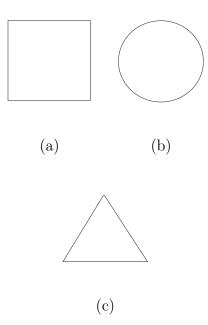


Figure 3.1. Two rows of graphics: (a) Square (b) Circle (c) Rectangle

Three figures across the page requires fairly small figures to fit within the Graduate School margins.

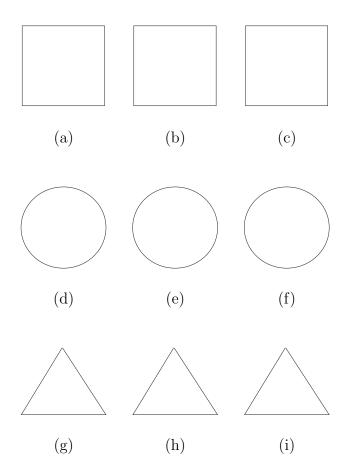


Figure 3.2. Three rows of graphics: (a)–(c) Squares. (d)–(f) Circles. (g)–(i) Ovals.

The verbatim environment can be useful when using data from a spreadsheet as is done below.

```
X,TRUE_SUR,MSE SIM,MSE ZHAO,MSE JIAN,MSE PZHAO,MSE PJIAN
      0.7520
              0.03864
                        0.01407
                                 0.01407
                                           0.01180
                                                     0.01223
 4.0
      0.7273
              0.04079
                        0.01675
                                 0.01675
                                           0.01479
                                                     0.01551
8.0
      0.7035
              0.04203
                        0.01923
                                 0.01923
                                           0.01675
                                                     0.01817
12.0
      0.6524
                        0.02157
                                 0.02135
                                           0.01932
                                                     0.02043
              0.04581
      0.6029
                                           0.02304
16.0
              0.05146
                        0.02345
                                 0.02266
                                                     0.02320
20.0
     0.5551
              0.05343
                        0.02498
                                 0.02393
                                           0.02627
                                                     0.02509
24.0
                        0.02677
                                           0.02936
      0.5089
              0.05449
                                 0.02453
                                                     0.02641
28.0
      0.4641
              0.05706
                        0.02901
                                 0.02442
                                           0.03315
                                                     0.02722
32.0
      0.4209
              0.05719
                        0.02910
                                 0.02341
                                           0.03558
                                                     0.02776
36.0
      0.3790
              0.05656
                        0.02974
                                 0.02229
                                           0.03745
                                                     0.02667
                        0.02940
40.0
      0.3385
              0.05518
                                 0.02119
                                           0.03864
                                                     0.02618
                        0.02989
44.0
      0.2994
              0.05344
                                           0.03928
                                                     0.02531
                                 0.02054
48.0
      0.2615
              0.04950
                        0.02803
                                 0.01906
                                           0.03855
                                                     0.02414
52.0
      0.2249
              0.04582
                        0.02712
                                 0.01812
                                           0.03849
                                                     0.02229
56.0
      0.1895
              0.04101
                        0.02454
                                 0.01578
                                           0.03632
                                                     0.01918
                                           0.03372
60.0
      0.1552
              0.03564
                        0.02282
                                 0.01315
                                                     0.01629
64.0
      0.1220
              0.03216
                        0.02124
                                 0.00997
                                           0.03188
                                                     0.01391
                                 0.00688
68.0
     0.0900
              0.02420
                        0.01730
                                           0.02551
                                                     0.01070
72.0
                        0.01254
                                           0.01811
      0.0590
              0.01592
                                 0.00363
                                                     0.00622
76.0
     0.0290
              0.00865
                        0.00838
                                 0.00110
                                           0.00886
                                                     0.00368
```

Figure 3.3. Use of verbatim environment

On the following page is an example of how to rotate text that is too long to fit within the horizontal margins that are required.

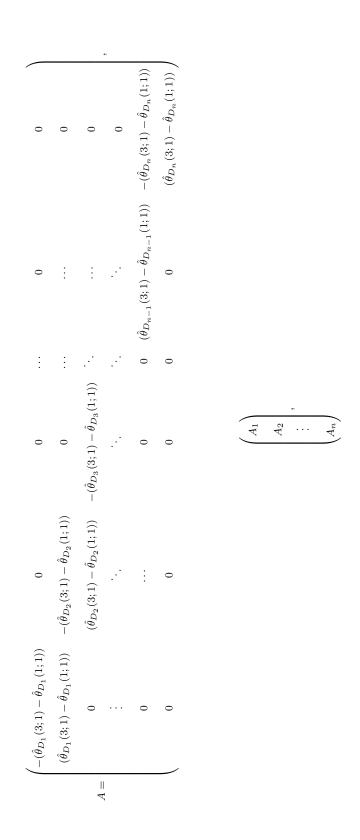


Figure 3.4. Matrix Rotated 90 degrees.

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Special Honors and Awards: (OMIT IF NONE)

Research Paper Title:

A Sample Research Paper on Aspects of Elementary Linear Algebra

Major Professor: Dr. J. Jones

Publications: (OMIT IF NONE)