Characterization of System on a Chip (SoC) Single Event Upset (SEU) Responses using SEU Data, Classical Reliability Models, and Space Environment Data

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Acronyms

- Combinatorial logic (CL)
- Commercial off the shelf (COTS)
- Complementary metal-oxide semiconductor (CMOS)
- Device under test (DUT)
- Edge-triggered flip-flops (DFFs)
- Error rate ($\lambda$)
- Error rate per bit ($\lambda_{bit}$)
- Error rate per system ($\lambda_{system}$)
- Field programmable gate array (FPGA)
- Global triple modular redundancy (GTMR)
- Hardware description language (HDL)
- Input – output (I/O)
- Intellectual Property (IP)
- Linear energy transfer (LET)
- Mean fluence to failure (MFTF)
- Mean time to failure (MTTF)
- Number of used bits (#Usedbits)
- Operational frequency (fs)
- Personal Computer (PC)

- Probability of configuration upsets ($P_{configuration}$)
- Probability of Functional Logic upsets ($P_{functionalLogic}$)
- Probability of single event functional interrupt ($P_{SEFI}$)
- Probability of system failure ($P_{system}$)
- Processor (PC)
- Radiation Effects and Analysis Group (REAG)
- Reliability over time ($R(t)$)
- Reliability over fluence ($R(\Phi)$)
- Single event effect (SEE)
- Single event functional interrupt (SEFI)
- Single event latch-up (SEL)
- Single event transient (SET)
- Single event upset (SEU)
- Single event upset cross-section ($\sigma_{SEU}$)
- Xilinx Virtex 5 field programmable gate array (V5)
- Xilinx Virtex 5 field programmable gate array radiation hardened (V5QV)
Problem Statement

• Conventional methods of applying single event upset (SEU) data to complex systems implemented in field programmable gate array (FPGA) devices need improvement.

• The problem boils down to extrapolation and application of SEU data to characterize system performance in radiation environments.
Abstract

• We are investigating the application of classical reliability performance metrics combined with standard SEU analysis data.
• We expect to relate SEU behavior to system performance requirements…
  – Example: The system is required to be 99.999% (5-nines) reliable within a given time window. Will the system’s SEU response meet mission requirements?
  – Our proposed methodology will provide better prediction of SEU responses in harsh radiation environments.
Background

FPGA SEU Susceptibility Measured in SEU Cross Section ($\sigma_{SEU}$)

- $\sigma_{SEUs}$ (per category) are calculated from SEE test and analysis.
- FPGAs vary and so do their SEU responses.
- Most believe the dominant $\sigma_{SEUs}$ are per bit (configuration or functional logic). However, global routes are also significant.

For functional logic, should $\sigma_{SEUs}$ be measured by bit???

$$P\left( f_s \right)_{system} \propto P_{Configuration} \sigma_{SEU} + P\left( f_s \right)_{functionalLogic} \sigma_{SEU} + P_{SEFI} \sigma_{SEU}$$

Sequential and Combinatorial logic (CL) in data path

Global Routes and Hidden Logic
Background
(Current Goal: Convert SEU cross-sections (\(\sigma_{SEU}:\) cm\(^2/(particles)\)) to error rates (\(\lambda\)) for complex systems)

- Perform SEU accelerated radiation testing across ions with different linear energy transfers (LETs) to calculate \(\sigma_{SEU}\)s per LET.
- **Bottom-Up approach** (transistor level):
  - Given \(\sigma_{SEU}\) (per bit) use an error rate calculator (such as CRÈME96) to obtain an error rate per bit (\(\lambda_{bit}\)).
  - Multiply \(\lambda_{bit}\) by the dominant number of used memory bits (\(#UsedBits\)) in the target design to attain a system error rate (\(\lambda_{system}\)).
- **Top-Down approach** (system level):
  - Given \(\sigma_{SEU}\) (per system) use an error rate calculator (such as CRÈME96) to obtain an error rate per bit (\(\lambda_{system}\)).
Technical Problems with Current Methods of Error Rate Calculation

- For submission to CRÈME96, $\sigma_{\text{SEU}}$ data (across LET) is fitted to a Weibull curve.
  - The two main parameters for curve fitting are a shape factor and a slope factor.
  - During the curve fitting process, a large amount of error can be introduced.
  - Consequently, it is possible for resultant error rates (for the same design) to vary by decades.

- Because of the error rate calculation process, $\sigma_{\text{SEU}}$ data is blended together and it is nearly impossible to hone in on the problem spots. This can become important for mitigation insertion.

![Top-down $\sigma_{\text{SEU}}$ Data versus LET](image-url)
Technical Problems with Bottom-Up Analysis Method (1)

- Multiplying each bit within a design by $\lambda_{\text{bit}}$ is not an efficient method of system error rate prediction.
  - Works well with memory structures…
  - but…complex systems do not operate like memories.
  - If an SEU affects a bit, and the bit is either inactive, disabled, or masked, a system malfunction might not occur.
    - Using the same multiplication factor across DFFs will produce extreme over-estimates.
    - To this date, there is no accurate method to predict DFF activity for complex systems.
    - Fault injection or simulation will not determine frequency of activity.

$$\lambda_{\text{system}} < \lambda_{\text{bit}} \times \#\text{UsedBits}$$
Technical Problems with Bottom-Up Analysis Method (2)

- There are a variety of components that are susceptible to SEUs (clocks, resets, combinatorial logic, flip-flops (DFFs, etc...)).
  - Various component susceptibilities are not accurately characterized at a per bit level.
  - Design topology makes a significant difference in susceptibility and is not characterized in error rate calculators (e.g., CREME96).

Error rates calculated at the transistor-bit level are estimated at too small of granularity for proper extrapolation to complex systems.
Let’s Not Reinvent The Wheel… A Proven Solution Can Be Found in Classical Reliability Analysis

- Classical reliability models have been used as a standard metric for complex system performance.
- The analysis provides a more in depth interpretation of system behavior over time by using system-level MTTF data for system performance metrics.

\[ R(t) = e^{-t/MTTF} \text{ or } R(t) = e^{-\lambda t} \]

Theory is already developed, proven, and should be in our hands!
Failure Rate ($\lambda(T)$) Bathtub Curve (Weibull Probability Density Function (PDF))

- Infant Mortality... error rate decreases with time
- Useful Life... Random errors (constant error rate)
- Wear Out Life... error rate increases with time

We will focus on the “Useful Life” of the bathtub curve for this analysis.
Mapping Classical Reliability Models from The Time Domain To The Fluence Domain

- The exponential model that relates reliability to MTTF assumes that during usefull-lifetime:
  - Failures are random.
  - Error rate is constant.
  - $\text{MTTF} = \frac{1}{\lambda}$.

- For a given LET (across fluence):
  - SEUs are random.
  - $\sigma_{\text{SEU}}$ is constant.
  - $\text{MFTF} = \frac{1}{\sigma_{\text{SEU}}}$.

- Hence, mapping from the time domain to the fluence domain (per LET) is straightforward:
  - $t \leftrightarrow \Phi$
  - MTTF $\leftrightarrow$ MFTF
  - $\lambda \leftrightarrow \sigma_{\text{SEU}}$
  
  $R(t) = e^{-t/\text{MTTF}} \quad \text{or} \quad R(t) = e^{-\lambda t}$

  \[ \text{Weibull slope} = 1 \ldots \text{exponential}. \]

  Parallel between time and fluence.

  $\sigma_{\text{SEU}} = \frac{\text{#errors}}{\text{fluence}}$

  $\lambda_{\text{system}} = \frac{\text{#errors}}{\text{time}}$
Creating Reliability Curves from $\sigma_{\text{SEU}}$

- $\sigma_{\text{SEU}}$ data is system level.
- A histogram of environment data is created. Bins are determined by LET values at each $\sigma_{\text{SEU}}$ data point.
- For each data point at a given LET, a combination of binned environment data and upper-bound $\sigma_{\text{SEU}}$ data are used to determine system reliability performance.
- A piecemeal approach is performed per data point to determine the weakest points of system performance.

Example

• Mission requirements:
  – The FPGA shall contain an embedded microprocessor.
  – Decision shall be made to select a Xilinx V5QV (approximately $80,000 per device) or a Xilinx V5 with embedded PowerPC (less than $2000.00) per device.
  – FPGA operation shall have reliability of 3-nines (99.9%) within a 10 minute window at Geosynchronous Equatorial Orbit (GEO).

• Proposed methodology:
  – Create a histogram of particle flux versus LET for a 10-minute window of time for your target environment.
  – Calculate MFTF per LET (obtain SEU data).
  – Graph R(\Phi) for a variety of LET values and their associated MFTFs. \( R(\Phi) = e^{\Phi/MFTF} \)
  – For selected ranges of LETs, use an upper bound of particle flux (number of particles/cm\(^2\)•10-minutes), to determine if the system will meet the mission’s reliability requirements.
Flux versus LET Histogram for A 10-minute Window

Geosynchronous Equatorial Orbit (GEO)
100-mils shielding

Flux (particles/(cm²*10-minutes))

LET Bins (MeVcm²/mg)

0 To 0.07 0.07 To 0.14 0.14 To 1.8 1.8 To 3.6 3.6 To 20 20 To 40 40 and over
MFTF versus LET for the Xilinx V5 MicroBlaze Soft Processor Core and the Xilinx V5QV embedded PowerPC Core

\[ \text{MFTF} = \frac{1}{\sigma_{\text{SEU}}} \]

Note: no system errors were observed for V5QV at \( \text{LET} < 3.6 \text{MeV cm}^2/\text{mg} \).

However, configuration bit errors were observed (design dependent).

We are focused on system performance.

Reliability across Fluence at LET=0.07MeV•cm²/mg And Below

- **V5QV:** no system errors were observed below LET=3.6MeV•cm²/mg. Total fluence > 5.0×10⁸ particles/cm².
- **PowerPC:**
  - No system errors were observed from an LET=0.07MeV•cm²/mg with total fluence = 1.0×10⁸ particles/cm².
  - Hence, at 0.07, we will assume an upper-bound MFTF = 1.0×10⁸ particles/cm².
  - More tests would increase the MFTF for this bin.

V5QV: MicroBlaze with Cache Enabled
V5: PowerPC
Reliability across Fluence up to LET=0.07 MeV\(\cdot\)cm\(^2\)/mg – Low Bound Analysis

Binned GEO Environment data shows approximately 3000 particles/(cm\(^2\)•10-minutes), in the range of 0.0MeV•cm\(^2\)/mg to 0.07MeV•cm\(^2\)/mg. We are using MFTF for 0.07MeV•cm\(^2\)/mg to upper bound this bin.

\[ R(\Phi) = e^{\Phi/1.0 \times 10^8} \]

Reliability at 3000 particles/(cm\(^2\)•10-minutes) > 99.99% for the PowerPC design implementation. “9’s” could be increased with more tests.
Reliability across Fluence up to LET=0.14 MeV·cm²/mg

Binned GEO Environment data shows approximately 11 particles/(cm²·10-minutes), in the range of 0.07 MeV·cm²/mg to 0.14 MeV·cm²/mg. We are using MFTF for 0.1 MeV·cm²/mg to upper bound this bin.

\[ R(\Phi) = e^{\Phi/5.0 \times 10^6} \]

Reliability at 5 particles/(cm²·10-minutes) > 99.999% for the V5QV PowerPC design implementation.

Reliability across Fluence up to LET=1.8 MeV•cm²/mg

Binned GEO Environment data shows approximately 9 particles/(cm²•10-minutes), in the range of 0.14MeV•cm²/mg to 1.8MeV•cm²/mg. We are using MFTF for 1.8MeV•cm²/mg to upper bound this bin.

We fall below 99.99% at approximately 6 particles/cm²!

Reliability at 9 particles/(cm²•10-minutes) > 99.9% for the PowerPC design implementation. This is the most susceptible bin for the system.
Reliability across Fluence up to LET=3.6MeV•cm²/mg

Binned GEO Environment data shows approximately 0.23 particles/(cm²•10-minutes), in the range of 1.8MeV•cm²/mg to 3.6MeV•cm²/mg.

Within this LET range, reliability at 0.23 particles/(cm²•10-minutes) > 99.999% for both design implementations.

\[
R(\Phi) = e^{\Phi/3.0 \times 10^6} \quad \text{V5QV: MFTF= 3.0\times 10^6} \\
R(\Phi) = e^{\Phi/1.2 \times 10^3} \quad \text{PowerPC: MFTF = 1.2\times 10^3}
\]
Binned GEO environment data shows approximately 0.07 particles/(cm\(^2\)•10-minutes), in the range of 3.6MeV•cm\(^2\)/mg to 40.0MeV•cm\(^2\)/mg.

Within this LET range, reliability at 0.07 particles/(cm\(^2\)•10-minutes) > 99.9% for both design implementations. We can refine by analyzing smaller bins.
Example Conclusion

- Using the proposed methodology, the commercial Xilinx V5 device will meet project requirements.
- In this case, the project is able to save money by selecting the significantly cheaper FPGA device and gain performance because of the embedded PowerPC.
Conclusions

• This study transforms proven classical reliability models into the SEU particle fluence domain. The intent is to better characterize SEU responses for complex systems.

• The method for reliability-model application is as follows:
  – SEU data are obtained as MFTF.
  – Reliability curves (in the fluence domain) are calculated using MFTF; and are analyzed with a piecemeal approach.
  – Environment data are then used to determine particle flux exposure within required windows of mission operation.

• The proposed method does not rely on data-fitting and hence removes a significant source of error.

• The proposed method provides information for highly SEU-susceptible scenarios; hence enabling a better choice of mitigation strategy.

• This is preliminary work. There is more to come.

This methodology expresses SEU behavior and response in terms that missions understand via classical reliability metrics.
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