

The Testing of Entropy Sources for Cryptography

Sarah Scheffler

Computer Security Division
Security Testing, Validation, & Management Group
Dr. Allen Roginsky

7 August, 2014

- 1 Background, definitions, and standards
- 2 Random-looking numbers with zero entropy
- 3 The collision test

- Everyone needs cryptography: personal, business, national, international
- Most cryptography heavily relies on random number generators (RNG)
- These RNGs get non-deterministic bits from an *entropy source*
- NIST has a draft publication for ensuring sufficient randomness of entropy sources

Definition: Entropy

$H(X)$ = the “uncertainty” in a random variable X

Examples:

- A fair coin toss has an entropy of 1 bit
- A set of eight fair coin tosses has an entropy of 8 bits
- What about a weighted coin toss?
 - ▶ *Worst-case* uncertainty

Definition: Min-entropy

If an adversary extracted *as much information* from this variable as possible, how much uncertainty remains?

$$H_{\infty}(X) = -\log p_{max}$$

- Min-entropy is the largest real number m such that all events occur with probability no greater than 2^{-m}
- A fair coin toss: $H_{\infty}(X) = -\log \frac{1}{2} = \mathbf{1 \text{ bit}}$
- A $\frac{3}{4}$ -weighted coin toss: $H_{\infty}(X) = -\log \frac{3}{4} \approx \mathbf{0.415 \text{ bits}}$

NIST DRAFT Special Publication 800-90B

Recommendation for the Entropy Sources Used for Random Bit Generation

Elaine Barker

John Kelsey

**Computer Security Division
Information Technology Laboratory**

The 800-90B statistical tests are split into two categories:

- IID tests
 - ▶ Shuffle the data to ensure IID
 - ▶ Calculate the min-entropy directly using $H_{\infty}(X) = -\log p_{max}$
- Non-IID tests
 - ▶ **Collision test**
 - ▶ Partial collection test
 - ▶ Markov test (map 6 bits)
 - ▶ Compression test
 - ▶ Frequency test

Definition: IID

A dataset that is IID (Independent and Identically Distributed) means that each sample is independent from each other sample, and that each sample uses the same distribution.

- Min-entropy is the “worst-case” measure of the uncertainty of a variable
- For the purposes of this talk, entropy and min-entropy are interchangeable
- IID means “independent and identically distributed.”
- NIST has statistical tests (800-90B) to evaluate entropy sources, sorted into IID and non-IID tests.
- My job: Evaluate the tests in 800-90B

Question

If an entropy source passes the statistical tests in 800-90B, is it secure?

Problem

We want a sequence of numbers that *looks* random, but we know each next outcome with 100% certainty.

Problem

We want a sequence of numbers that *looks* random, but we know each next outcome with 100% certainty.

Soultion

Test digits of irrational numbers like π .

Constants chosen

- π
- $\zeta(3)$
- $\log 2$

The entropy of π

```
Read in file ../../general-bbp/pi/piBin/piBin_8, 1250000 bytes long.
Dataset: 1250000 8-bit symbols.
Output symbol values: min = 0, max = 255
collision test not run:
mu_bar = 20.7708, max valid value for this test and model = 20.7261
- collision test *not valid* for this data set.
- Partial collection test : p(max) = 0.00609112, min-entropy = 7.35908
- Markov test (map 6 bits): p(max) = 1.05961e-223, min-entropy = 5.78677
- Compression test      : p(max) = 0.00686646, min-entropy = 7.18622
- Frequency test        : p(max) = 0.0041008, min-entropy = 7.58853
```

```
*****
** min-entropy = 5.78677 7.18622
*****
```

```
Read in file ../../general-bbp/pi/picopy_8_8, 1250000 bytes long.
Dataset: 1250000 8-bit symbols.
Output symbol values: min = 0, max = 255
```

```
Compression Test:      Result = passed      Over/under Test:      Result = passed
Excursion Test:        Result = passed      Directional runs Test: Result = passed
Covariance Test:       Result = passed      Collision Test:        Result = passed
** Passed iid shuffle tests
Chi square independence score = 65169.2, degrees of freedom = 65280, cut-off = 66402.2
** Passed chi-square independence test
Chi square stability   score = 2358.01, degrees of freedom = 2313 cut-off = 2528.88
** Passed chi-square stability test
```

```
*****
min-entropy for ../../general-bbp/pi/picopy_8_8 = 7.88435
*****
```

The entropy of $\zeta(3)$

```
Read in file ../../../../zeta3-bbp/zeta3Bin/zeta3Bin_8, 1250000 bytes long.
Dataset: 1250000 8-bit symbols.
Output symbol values: min = 0, max = 255
- Collision test          : p(max) = 0.00997925, min-entropy = 6.64685
- Partial collection test : p(max) = 0.00644398, min-entropy = 7.27783
- Markov test (map 6 bits): p(max) = 2.24652e-223, min-entropy = 5.7783
- Compression test       : p(max) = 0.00830078, min-entropy = 6.91254
- Frequency test         : p(max) = 0.0040824, min-entropy = 7.59365
```

```
*****
* min-entropy = 5.7783 6.64685
*****
```

```
Read in file ../../../../zeta3-bbp/zeta3Bin_8_8, 1250000 bytes long.
Dataset: 1250000 8-bit symbols.
Output symbol values: min = 0, max = 255
```

```
Compression Test:      Result = passed      over/under Test:      Result = passed
Excursion Test:        Result = passed      Directional runs Test: Result = passed
Covariance Test:       Result = passed      Collision Test:        Result = passed
** passed iid shuffle tests
Chi square independence score = 65424.1, degrees of freedom = 65280, cut-off = 66402.2
** passed chi-square independence test
Chi square stability   score = 2263.29, degrees of freedom = 2313 cut-off = 2528.88
** passed chi-square stability test
```

```
*****
min-entropy for ../../../../zeta3-bbp/zeta3Bin_8_8 = 7.89074
*****
```

The entropy of log 2

```
Read in file ../../general-bbp/log2/log2Bin, 1250000 bytes long.
Dataset: 1250000 8-bit symbols.
Output symbol values: min = 0, max = 255
- Collision test          : p(max) = 0.00619507, min-entropy = 7.33466
- Partial collection test : p(max) = 0.00695324, min-entropy = 7.1681
- Markov test (map 6 bits): p(max) = 9.35661e-224, min-entropy = 5.78817
- Compression test       : p(max) = 0.00756836, min-entropy = 7.0458
- Frequency test          : p(max) = 0.0041352, min-entropy = 7.57901
```

```
***
** min-entropy = 5.78817 7.0458
***
```

```
Read in file ../../general-bbp/log2/log2Bin, 1250000 bytes long.
Dataset: 1250000 8-bit symbols.
Output symbol values: min = 0, max = 255
```

```
Compression Test:      Result = passed      Over/under Test:      Result = passed
Excursion Test:        Result = passed      Directional runs Test: Result = passed
Covariance Test:       Result = passed      Collision Test:        Result = passed
** Passed iid shuffle tests
Chi square independence score = 64754.7, degrees of freedom = 65280, cut-off = 66402.2
** Passed chi-square independence test
Chi square stability   score = 2186.42, degrees of freedom = 2313 cut-off = 2528.88
** Passed chi-square stability test
```

```
***
min-entropy for ../../general-bbp/log2/log2Bin = 7.87249
***
```

What have we learned?

What have we learned?

- A constant is an entropy source with *zero* entropy.

What have we learned?

- A constant is an entropy source with *zero* entropy.
- But, these constants pass all our tests for entropy.

What have we learned?

- A constant is an entropy source with *zero* entropy.
- But, these constants pass all our tests for entropy.
- Conceptually, there is *no way* to tell the difference between random and nonrandom entropy sources using only the output.

What have we learned?

- A constant is an entropy source with *zero* entropy.
- But, these constants pass all our tests for entropy.
- Conceptually, there is *no way* to tell the difference between random and nonrandom entropy sources using only the output.
- **There must be a qualitative as well as quantitative assessment of entropy sources.**

Back to the pi result...

```
Read in file ../../general-bbp/pi/piBin/piBin_8, 1250000 bytes long.  
Dataset: 1250000 8-bit symbols.  
Output symbol values: min = 0, max = 255
```

```
collision test not run:  
mu_bar = 20.7708, max valid value for this test and model = 20.7261  
- collision test *not valid* for this data set.  
- Partial collection test : p(max) = 0.00609112, min-entropy = 7.35908  
- Markov test (map 6 bits): p(max) = 1.05961e-223, min-entropy = 5.78677  
- Compression test       : p(max) = 0.00686646, min-entropy = 7.18622  
- Frequency test         : p(max) = 0.0041008, min-entropy = 7.58853
```

```
*****
```

```
** min-entropy = 5.78677
```

```
*****
```

The Collision Test

Measures the min-entropy by observing time between “collisions”
(repeated outputs)

- Data set: BACBCBACCAABAABCBC
- Marked collisions: BACB CBAC CAA BAA BCB C
- Collision difference times: 4 4 3 3 3
- Collision statistic: $\text{avg}(4, 4, 3, 3, 3)$

- From a data set $\{X_s\}$, define a sequence of collision times $\{T_i\}$, where $T_0 = 0$, and

$$T_i = \min \{j > T_{i-1} : \exists m \in (T_{i-1}, j) \text{ such that } X_j = X_m\}$$

Collision Statistic

S_k = average of differences of collision times

$$\begin{aligned} S_k &= \frac{1}{k} \sum_{i=1}^k (T_i - T_{i-1}) \\ &= \frac{1}{k} ((T_1 - T_0) + (T_2 - T_1) + \cdots + (T_k - T_{k-1})) \\ &= \boxed{\frac{T_k}{k}} \end{aligned}$$

Collision Statistic

$$S_k = \frac{T_k}{k}$$

- Calculate expected value for a probability distribution $\mathbf{p}$:

$$\begin{aligned}\mathbb{E}_{\mathbf{p}}[S_k] &= \mathbb{E}_{\mathbf{p}}[T_k/k] \\ &= \frac{1}{k} \mathbb{E}_{\mathbf{p}}[T_k] \\ &= \mathbb{E}_{\mathbf{p}}[T_1] \\ &= \sum_{i=1}^{n+1} P_i\end{aligned}$$

(P_i = probability that no collisions have occurred after i outputs)

Collision Statistic

$$S_k = \frac{T_k}{k}$$

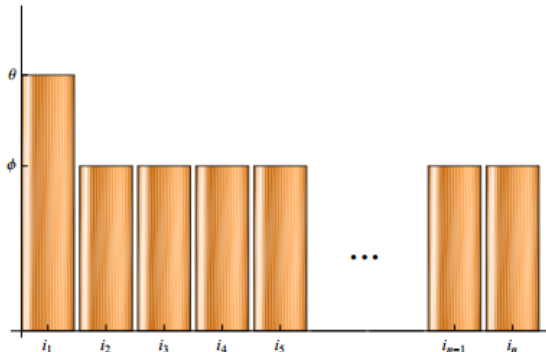
- Calculate expected value for a probability distribution $\mathbf{p}$:

$$\begin{aligned}\mathbb{E}_{\mathbf{p}}[S_k] &= \mathbb{E}_{\mathbf{p}}[T_k/k] \\ &= \frac{1}{k} \mathbb{E}_{\mathbf{p}}[T_k] \\ &= \mathbb{E}_{\mathbf{p}}[T_1] \quad (\text{requires IID}) \\ &= \sum_{i=1}^{n+1} P_i\end{aligned}$$

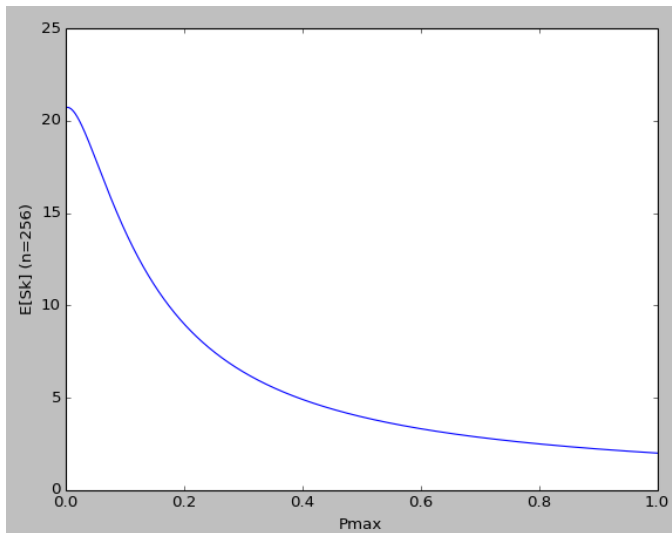
(P_i = probability that no collisions have occurred after i outputs)

Near-uniform distribution family

$$\mathbf{p}_\theta[Z = i] = \begin{cases} \theta & i = i_1, \\ \frac{1-\theta}{n-1} & \text{otherwise.} \end{cases}$$

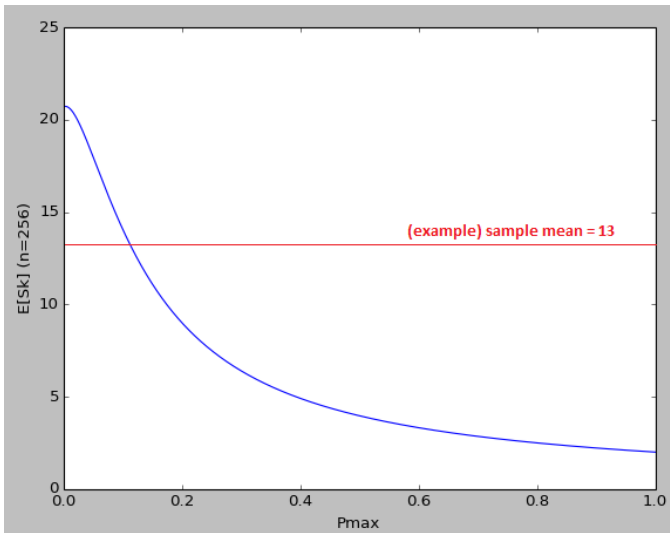


Steps of the Collision Test



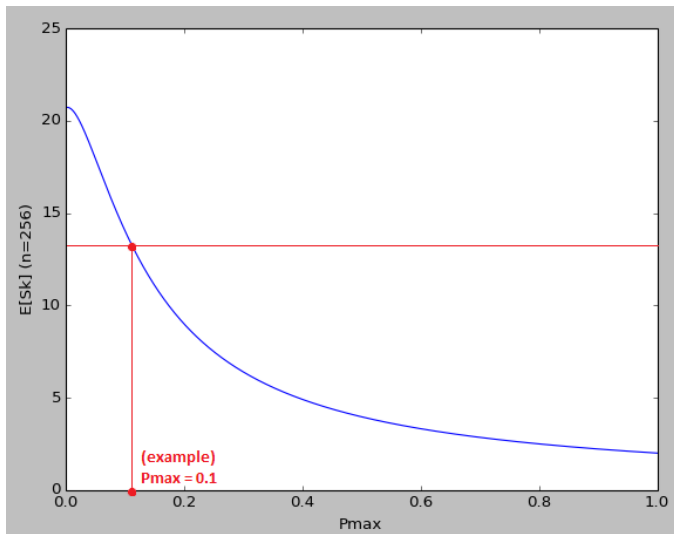
- 1 Get a family of near-uniform expected values

Steps of the Collision Test



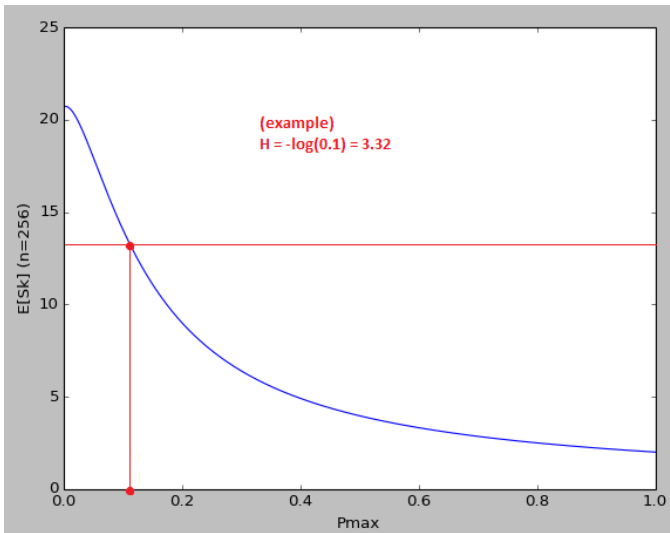
- 2 Compare sample mean to near-uniform expected values

Steps of the Collision Test



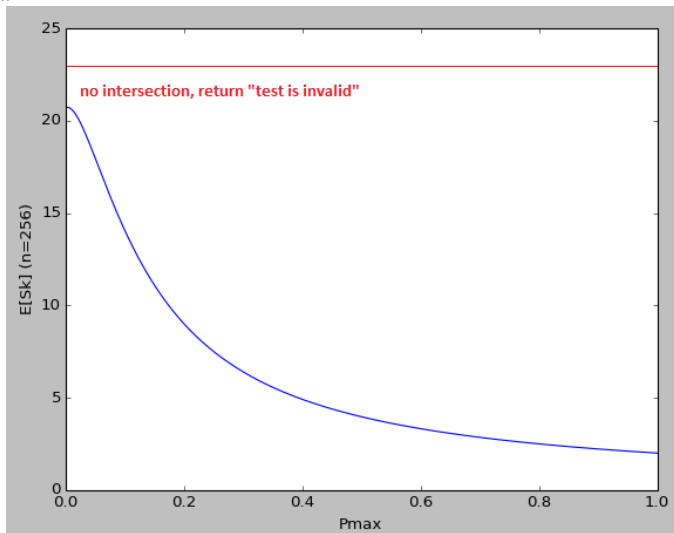
- ③ Get the p_{max} that corresponds to this expected value

Steps of the Collision Test



1 Calculate min-entropy directly

Why is the test invalid?



- There is no intersection, so the test cannot calculate the min-entropy

Collision Statistic

$$S_k = \frac{T_k}{k}$$

- Calculate expected value for a probability distribution $\mathbf{p}$:

$$\begin{aligned}\mathbb{E}_{\mathbf{p}}[S_k] &= \mathbb{E}_{\mathbf{p}}[T_k/k] \\ &= \frac{1}{k} \mathbb{E}_{\mathbf{p}}[T_k] \\ &= \mathbb{E}_{\mathbf{p}}[T_1] \quad (\text{requires IID}) \\ &= \sum_{i=1}^{n+1} P_i\end{aligned}$$

(P_i = probability that no collisions have occurred after i outputs)

Conclusions

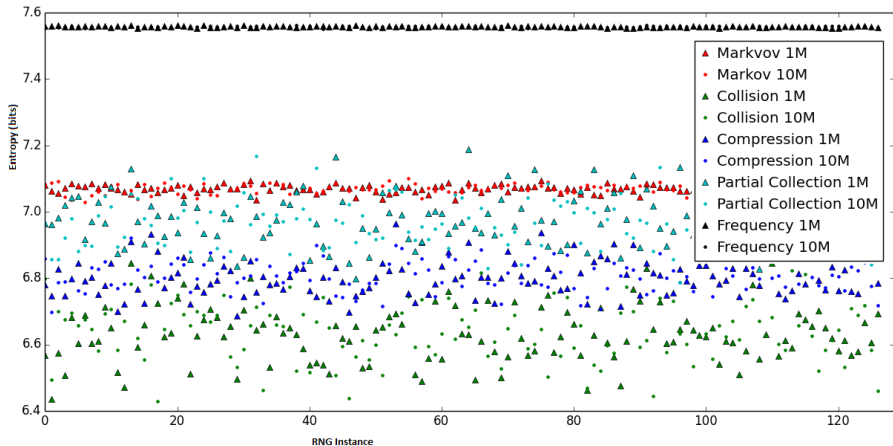
- Even when the Collision Test “succeeds,” it may be overestimating the expected value, which would cause a false increase in the entropy measurement.

- Even when the Collision Test “succeeds,” it may be overestimating the expected value, which would cause a false increase in the entropy measurement.
- **The Collision Test is being run on non-IID data, but it requires its data to be IID.**

- Even when the Collision Test “succeeds,” it may be overestimating the expected value, which would cause a false increase in the entropy measurement.
- **The Collision Test is being run on non-IID data, but it requires its data to be IID.**
- The Partial Collection Test and the Compression Test also assume IID, but are not in the IID section of 800-90B.

- Even when the Collision Test “succeeds,” it may be overestimating the expected value, which would cause a false increase in the entropy measurement.
- **The Collision Test is being run on non-IID data, but it requires its data to be IID.**
- The Partial Collection Test and the Compression Test also assume IID, but are not in the IID section of 800-90B.
- But when the data are IID, why not just measure the min-entropy directly?

“Non-IID” tests in 800-90B



- My advisor, Dr. Allen Roginsky, for giving me this project, for providing help and guidance, and for having really cool conversations
- Dr. Tim Hall, for being the source of many insights, for showing me the Python code, and for a lot of support
- Fellow SURF intern Bobby Staples, for helping me figure out the 90B's, and sending me that scatterplot
- The SURF directors, especially Dr. David Griffith, for organizing many events and helping me through some logistics
- Dr. Isabel Beichl, for providing some neat opportunities and conversations
- The Crypto Cave, for being awesome and fun friends
- Anyone not on this list