

# Deciding a sequence is "random"

*(generated by a fair coin)*

Fair coin

Biased coin:  $p(0)=3/4$

$$p_{\{2,3\}} = 0.62$$

11000	11100
10100	11010
10010	11001
10001	01110
01100	01101
01010	01011
01001	00111
00110	10110
00101	10101
....	....

$$p_{\{0,5\}} = 0.06$$

00000  
11111

$$p_{\{1,4\}} = 0.312$$

11110  
11101  
11011  
10111  
01111  
....

$$p_{\{2,3\}} = 0.36$$

11000	11100
10100	11010
10010	11001
10001	01110
01100	01101
01010	01011
01001	00111
00110	10110
00101	10101
....	....

$$p_{\{0,5\}} = 0.23$$

00000  
11111

$$p_{\{1,4\}} = 0.41$$

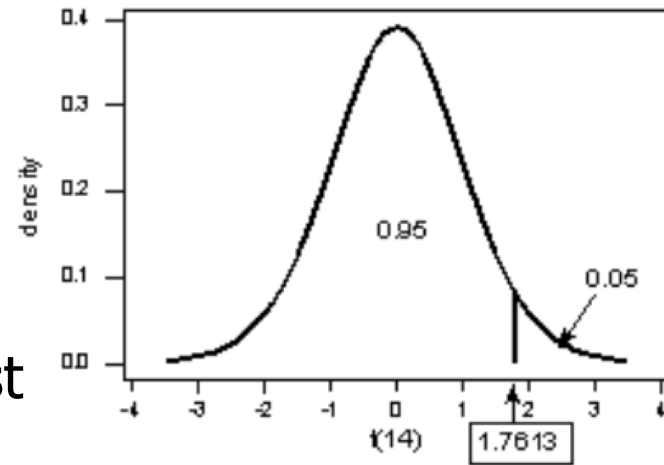
11110  
11101  
11011  
10111  
01111  
....

# Hypothesis testing & Statistical tests

- Given the sequence 00100, which of the two hypotheses hold:
- **Null Hypothesis  $H_0$ :** "the sequence is generated by a fair coin (the generator is a good RNG)"
- **Alternative Hypothesis:  $H_a$**  is "the sequence is not generated by a good generator"
- Use a **statistical test** to decide:
- Statistical tests rely on **test statistics**:
  - A "test statistics" is a value that can be calculated for a sequence and has a known distribution under  $H_0$

# Statistical tests

- A “test statistics” that,
  - can be calculated for a sequence
  - has a known distribution under  $H_0$
- Choose a significance level  $\alpha$ 
  - Probability of Type I error
- Using the known distribution of the test statistic, calculate the P-value
- If  $P\text{-value} < \alpha$ , then the null hypothesis is rejected
  - the sequence appears to be random.
- If  $P\text{-value} \geq \alpha$ , the null hypothesis is accepted; i.e.,



# NIST Frequency (Monobit) Test

- Calculate test statistics and P-value
- Choose a significance level
- Decide by comparing the p-value with the significance level

For example, if  $\varepsilon = 1011010101$ , then  $n=10$  and  $S_n = 1 + (-1) + 1 + 1 + (-1) + 1 + (-1) + 1 + (-1) + 1 = 2$ .

(2) Compute the test statistic  $s_{obs} = \frac{|S_n|}{\sqrt{n}}$

For the example in this section,  $s_{obs} = \frac{|2|}{\sqrt{10}} = .632455532$ .

(3) Compute  $P\text{-value} = \text{erfc}\left(\frac{s_{obs}}{\sqrt{2}}\right)$ , where *erfc* is the complementary error function as defined in Section 5.5.3.

**Complementary Error Function**

$$\text{erfc}(z) = \frac{2}{\sqrt{\pi}} \int_z^{\infty} e^{-u^2} du$$

For the example in this section,  $P\text{-value} = \text{erfc}\left(\frac{.632455532}{\sqrt{2}}\right) = 0.527089$ .

### 2.1.5 Decision Rule (at the 1% Level)

If the computed  $P\text{-value}$  is  $< 0.01$ , then conclude that the sequence is non-random. Otherwise, conclude that the sequence is random.

### 2.1.6 Conclusion and Interpretation of Results

Since the  $P\text{-value}$  obtained in step 3 of Section 2.1.4 is  $\geq 0.01$  (i.e.,  $P\text{-value} = 0.527089$ ), the conclusion is that the sequence is random.

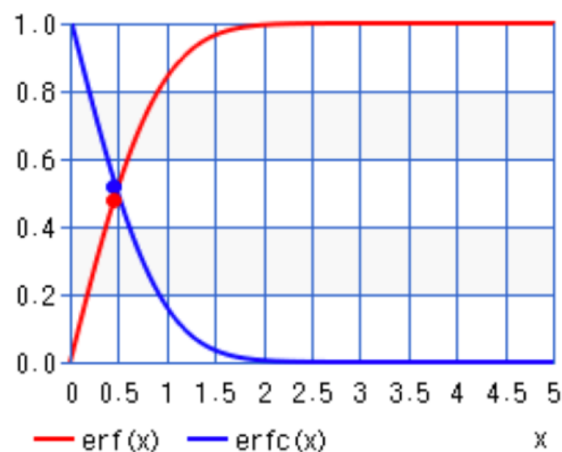
For example, if  $\varepsilon = 10$   
 $(-1) + 1 = 2$ .



## Error function Calculator

[Home](#) / [Special Function](#) / [Error function](#)

**Calculates the error function  $\text{erf}(x)$  and complementary error function  $\text{erfc}(x)$ .**



x

real number

**Execute**

**Clear**

**Store/Read**

**Print**

22digit



Error function	result
● $\text{erf}(x)$	0.4767973472542054181176
● $\text{erfc}(x)$	0.5232026527457945818824

(2) Compute the test statis

For the example in this

(3) Compute  $P\text{-value} = \text{erf}$

Section 5.5.3.

For the example in this

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# NIST Statistical Test Suite

- A statistical framework:
- **Randomness is a probabilistic property**; that is, the properties of a random sequence can be characterized and described in terms of probability. The likely outcome of statistical tests, when applied to a truly random sequence, is known a priori and can be described in probabilistic terms. There are an **infinite number of possible statistical tests**, each assessing the presence or absence of a “pattern” which, if detected, would indicate that the sequence is nonrandom. Because there are so many tests for judging whether a sequence is random or not, **no specific finite set of tests is deemed “complete.”** In addition, the results of statistical testing must be interpreted with some care and caution to avoid incorrect conclusions about a specific generator (see Section 4).

# Decision using Hypothesis Testing

- A statistical test is formulated to test a specific null hypothesis ( $H_0$ ). For the purpose of this document, the null hypothesis under test is that the sequence being tested is random. Associated with this null hypothesis is the alternative hypothesis ( $H_a$ ), which, for this document, is that the sequence is not random.
- For each applied test, a decision or conclusion is derived that accepts or rejects the null hypothesis, i.e., whether the generator is (or is not) producing random values, based on the sequence that was produced."

Statistical hypothesis tests define a procedure that controls (fixes) the probability of incorrectly deciding that a default position (null hypothesis) is incorrect. The procedure is based on how likely it would be for a set of observations to occur if the null hypothesis were true.



# Summary

- Random number generation
  - PRNG
    - Statistical uniformity
    - Unpredictability
      - Computational
  - TRNG
- Tests for evaluating randomness of a string

