



julia

Statistics with Julia

- ▶ Its all very well generating myriad statistics characterising your data.
- ▶ How do you know whether or not those statistics are telling you something interesting? Hypothesis Tests.
- ▶ To that end, well be looking at the HypothesisTests package today.

Statistics with Julia

```
Julia> using HypothesisTests
```

```
julia> using Distributions
```

```
julia> srand(357)
```

```
julia> x1 = rand(Normal(), 1000);
```

```
julia> x2 = rand(Normal(0.5, 1), 1000);
```

```
julia> # 25% success rate on samples of size 100
```

```
julia> x3 = rand(Binomial(100, 0.25), 1000);
```

```
julia> # 50% success rate on samples of size 50
```

```
julia> x4 = rand(Binomial(50, 0.50), 1000);
```

```
julia> x5 = rand(Bernoulli(0.25), 100) .== 1;
```

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- ▶ We'll apply a one sample t-test to x_1 and x_2 . The output below indicates that x_2 has a mean which differs significantly from zero while x_1 does not.
- ▶ This is consistent with our expectations based on the way that these data were generated.
- ▶ I'm impressed by the level of detail in the output from `OneSampleTTest()`: different aspects of the test are neatly broken down into sections (population, test summary and details) and there is automated high level interpretation of the test results.

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```
julia> t1 = OneSampleTTest(x1)
One sample t-test
-----
Population details:
parameter of interest:    Mean
value under h_0:          0
point estimate:           -0.013027816861268473
95% confidence interval:  (-0.07587776077157478,0.049822

Test summary:
outcome with 95% confidence: fail to reject h_0
two-sided p-value:        0.6842692696393744 (not si
```

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Details:

```
number of observations:    1000
t-statistic:               -0.40676289562651996
degrees of freedom:       999
empirical standard error: 0.03202803648352013
julia> t2 = OneSampleTTest(x2)
One sample t-test
```

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Population details:

parameter of interest:	Mean
value under h_0 :	0
point estimate:	0.507852246706
95% confidence interval:	(0.4468203, 0.568884)

Test summary:

outcome with 95% confidence:	reject h_0
two-sided p-value:	$2.62561601163e-53$

Details:

number of observations:	1000
t-statistic:	16.32883382693
degrees of freedom:	999
empirical standard error:	0.03110156255427

- ▶ Using `pvalue()` we can further interrogate the p-values generated by these tests.
- ▶ The values reported in the output above are for the two-sided test, but we can look specifically at values associated with either the left- or right tails of the distribution.
- ▶ This makes the outcome of the test a lot more specific.

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```
julia> pvalue(t1)  
0.6842692696393744
```

```
julia> pvalue(t2)  
2.6256160116367554e-53
```

```
julia> pvalue(t2, tail = :left)                                     #  
1.0
```

```
julia> pvalue(t2, tail = :right)  
1.3128080058183777e-53
```

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The associated confidence intervals are also readily accessible. We can choose between two-sided or left/right one-sided intervals as well as change the significance level.

```
julia> ci(t2, tail = :both)
(0.44682036100064954, 0.5688841324132342)
```

```
julia> ci(t2, tail = :left)
(-Inf, 0.5590572480083876)
```

```
julia> ci(t2, 0.01, tail = :right)
(0.43538291818831604, Inf)
```

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- ▶ As a second (and final) example we'll look at `BinomialTest()`.
- ▶ There are various ways to call this function. First, without looking at any particular data, we'll check whether 25 successes from 100 samples is inconsistent with a 25% success rate (obviously not and, as a result, we fail to reject this hypothesis).

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```
julia> BinomialTest(25, 100, 0.25)
Binomial test
-----
Population details:
parameter of interest:    Probability of success
value under h_0:         0.25
point estimate:          0.25
95% confidence interval: (0.16877973809,0.3465524957)
```

```
Test summary:
outcome with 95% confidence: fail to reject h_0
two-sided p-value:          1.0 (not significant)
```

```
Details:
number of observations: 100
number of successes:    25
```

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Next we'll see whether the Bernoulli samples in `x5` provide contradictory evidence to an assumed 50% success rate (based on the way that `x5` was generated we are not surprised to find an infinitesimal p-value and the hypothesis is soundly rejected).

```
julia> BinomialTest(x5, 0.5)
```

```
Binomial test
```

```
-----
```

```
Population details:
```

```
parameter of interest:    Probability of success
```

```
value under h_0:          0.5
```

```
point estimate:           0.18
```

```
95% confidence interval: (0.11031122915326055,0.26947708596
```

```
Test summary:
```

```
outcome with 95% confidence: reject h_0
```

```
two-sided p-value:          6.147806615048005e-11 (extreme
```

```
Details:
```

```
number of observations: 100
```

```
number of successes:      18
```

- ▶ There are a number of other tests available in this package, including a range of non-parametric tests which I have not even mentioned above.
- ▶ Certainly HypothesisTests should cover most of the bases for statistical inference. For more information, read the extensive documentation.