

# Hypothesis Testing:

Power, Type-I Error Rate, and Sample  
Size Calculation

# (Neyman-Pearson) Hypothesis Testing Framework

- A statistical model  $p(\mathbf{y}|\boldsymbol{\theta})$
- Model parameters  $\boldsymbol{\theta}=[\boldsymbol{\theta}_1, \boldsymbol{\theta}_2]$
- Hypotheses:  $H_0: \boldsymbol{\theta}_1=0$  ;  $H_A: \boldsymbol{\theta}_1 \neq 0$  [e.g., a regression coef. =0]
- Test statistic:  $S(\mathbf{y})$  [e.g., t-stat, F-stat, Chi-sq. stat]
- Decision rule: Reject  $H_0$  if  $S(\mathbf{y}) > \tau$  [e.g., reject if | t-stat | > 1.96]
- Possible cases

	Don't reject	Reject
$H_0$	OK	Type-I error
$H_A$	Type-II error	OK

- The test-statistic,  $S(\mathbf{y})$ , is a function of the data; thus it is random
- It has a sampling distribution (which describes how it varies over conceptual repeated sampling).

# Type-I error rate & Power

## Type-I error rate:

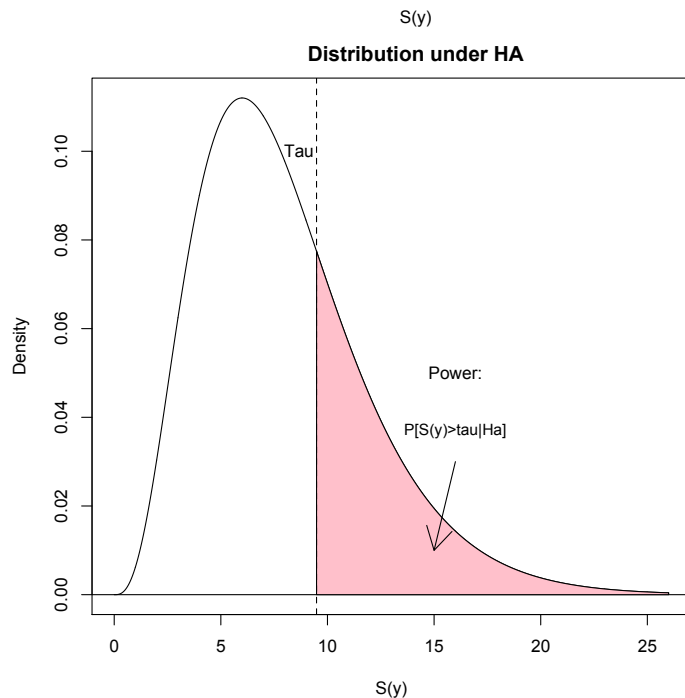
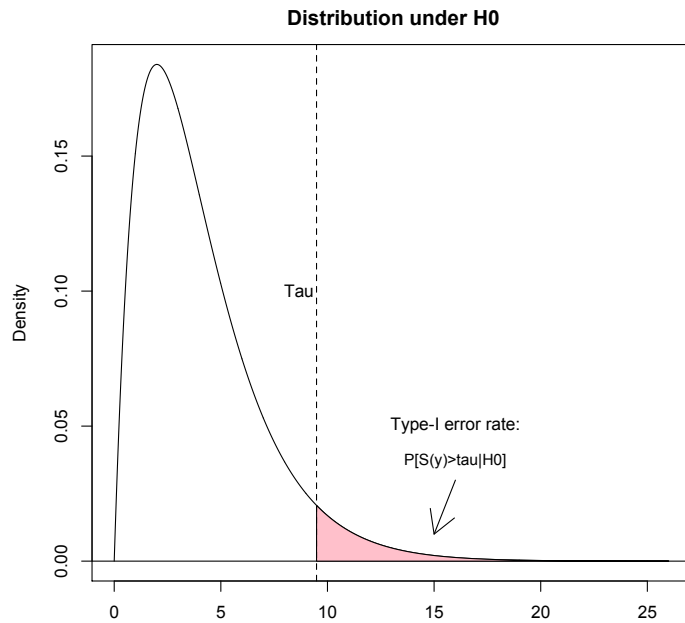
- Definition: Probability of rejecting, given that the null holds.
- We choose tau to control Type-I error rate low.
- P-values are estimates of type-I error rate

## Power:

- Definition: Probability of rejecting, given that  $H_a$  holds.
- The distribution of the test statistic under  $H_0$  is shifted (in this example, shifted towards the right)
- The area above the threshold is the power of the test.
- What factors affect power: Sample size, effect size, & and the test used.

## Optimal decision rule:

- There are trade-offs between Type-I error rate and Power
- If we move tau “right” we reduce type-I error rate but this also reduces power.
- Approach: chose a rule that minimizes Type-II error while controlling Type-I error rate smaller than a given threshold (significance level).
- To achieve this we reject if p-value is smaller than the desired significance level.



# Estimation of Type-I error rate & Power

## Analytical methods:

- In some cases, based on either assumptions or asymptotic theory, we can have a good guess of the distribution of the test statistic under  $H_0$  and  $H_a$ .
- In these cases we can estimate power and type-I error rate analytically

## Monte Carlo Simulations:

- However, in many cases the distribution of the test statistic under either  $H_0$  or  $H_a$  may not have a closed form
- In these cases we can estimate power and type-I error rate using simulations

## Sample size calculation

We can use results from power analysis to estimate the minimum sample size that would be required for an experiment to achieve certain power.