

Lecture 30 The Behaviour of The p-value

BIO210 Biostatistics

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SUSTech · SCHOOL OF
LIFE SCIENCES

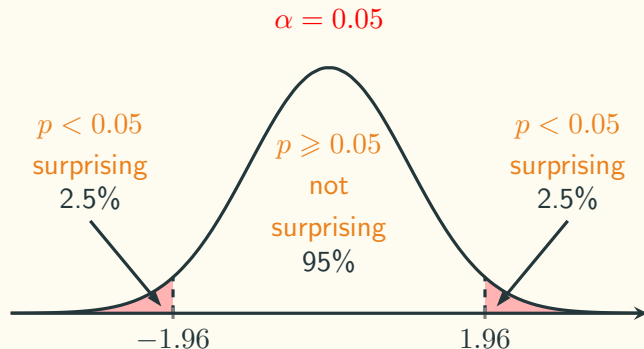
Why p-values Are Successful In Science

In some sense it offers a first line of defense against being fooled by randomness, separating signal from noise.

– Benjamini, 2016

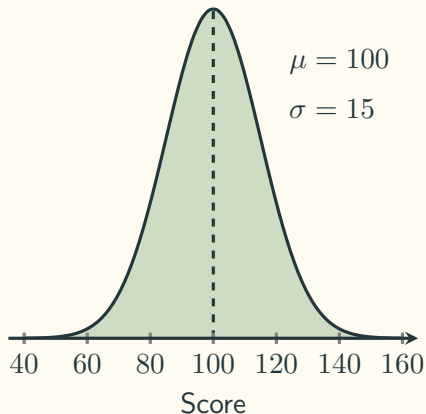
Why p-values Are Successful In Science

- $p\text{-value} = \mathbb{P}(\text{observed data or more extreme} \mid H_0 \text{ is true})$: How surprising the data is, assuming there is no effect?
- $p\text{-value}$ calculation: using the distribution of the test statistics, which is based on the sampling distribution.

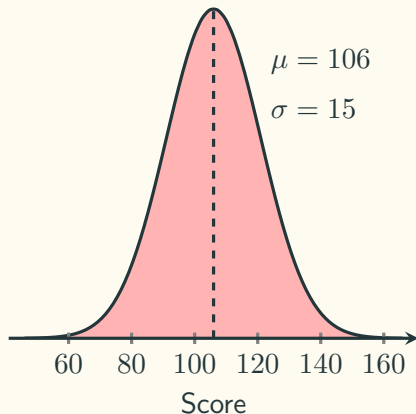


Test Scores In Math Exams - Two Populations

Students (≤ 5 hours per week)



Students (> 5 hours per week)



Test Scores In Math Exams - Hypothesis Testing

Question: are the mean scores the same between students who study ≤ 5 hours a week and those who study > 5 hours a week?

Simulation Setup



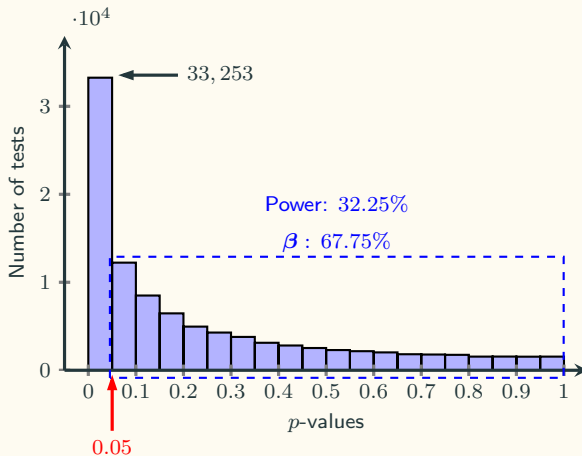
```
import numpy as np
from scipy.stats import norm
from scipy.stats import ttest_ind as tt

np.random.seed(42)                                # set seed for reproducible results
pop1 = norm(loc=100, scale=15)                     # set up population 1 (study hours < 5)
pop2 = norm(loc=106, scale=15)                     # set up population 2 (study hours > 5)
```

Two-Sample t -tests



```
pvals = np.zeros((100000,))  
  
for i in range(100000):  
    s1 = pop1.rvs(size=30)  
    s2 = pop2.rvs(size=30)  
    ts, p = tt(s1, s2, equal_var = True)  
    pvals[i] = p
```



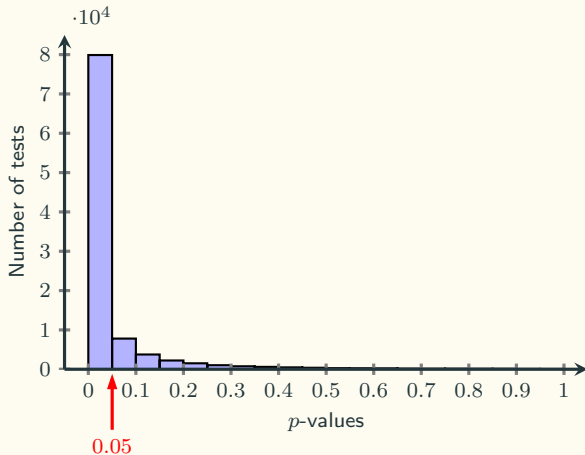
Distribution of p -values

We want to increase our power to 80%: $n = \left[\frac{(1.96 + 0.842) \times \sqrt{450}}{106 - 100} \right]^2 = 99$



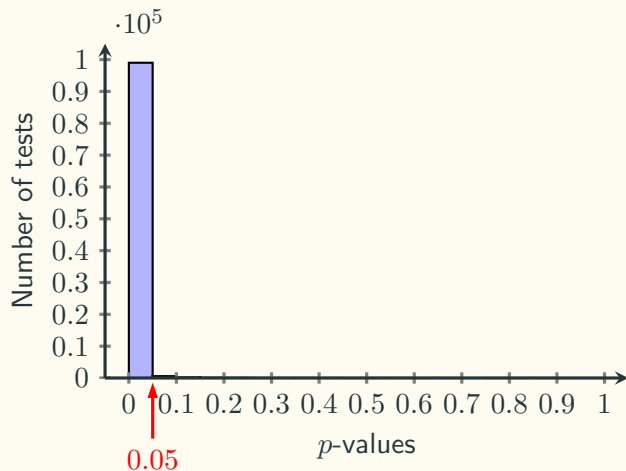
```
pvals = np.zeros((100000,))

for i in range(100000):
    s1 = pop1.rvs(size=99)
    s2 = pop2.rvs(size=99)
    ts, p = tt(s1, s2, equal_var = True)
    pvals[i] = p
```

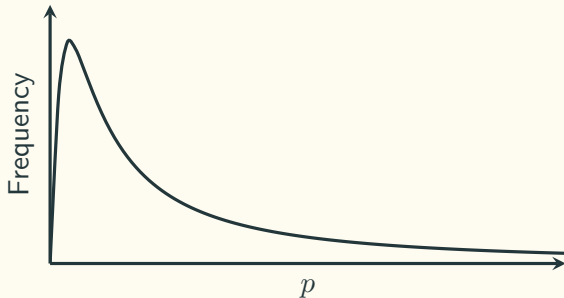


Distribution of p -values

Sample size: $n = 231$



Distribution of p -values When H_1 Is True

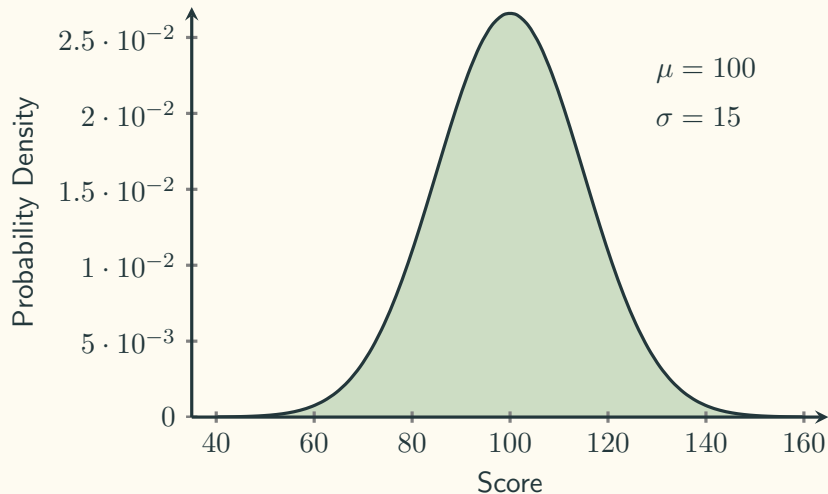


When H_1 is true, the distribution of p -values are skewed to the right, and the shape depends on the power.

What is the distribution of p -values when H_0 is true ?

Test Scores In Math Exams - Drinking Coke

Students (≥ 355 mL Coke per week)

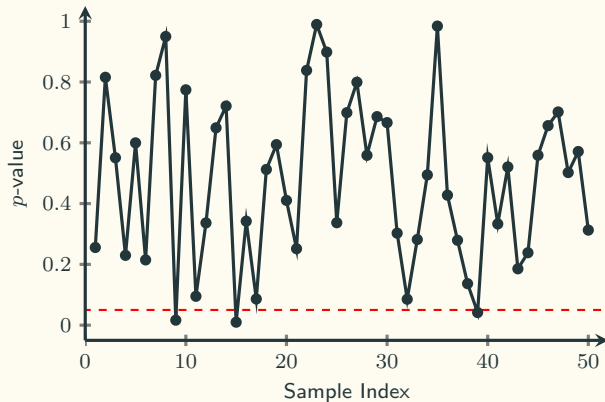


Take samples of
sizes $n = 100$

p -value Fluctuation

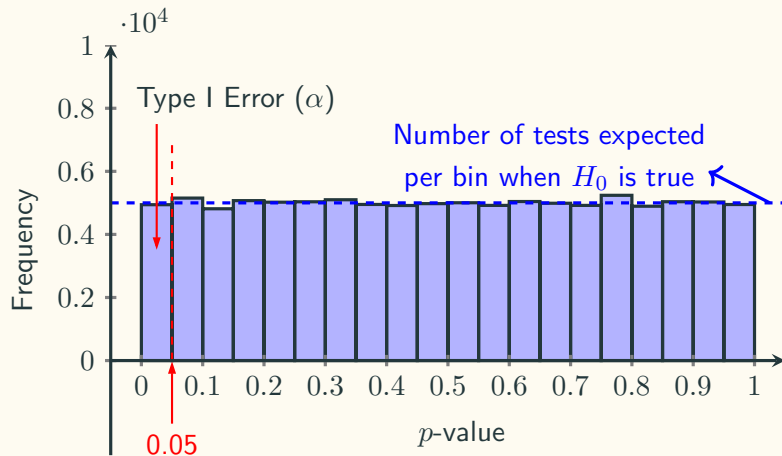


```
pvals = np.zeros((100000,))  
  
for i in range(100000):  
    s1 = pop1.rvs(size=100)  
    s2 = pop1.rvs(size=100)  
    ts, p = tt(s1, s2, equal_var = True)  
    pvals[i] = p
```

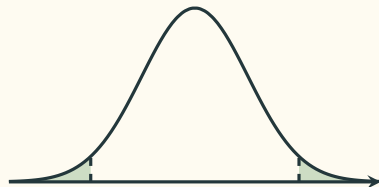


----- $p = 0.05$

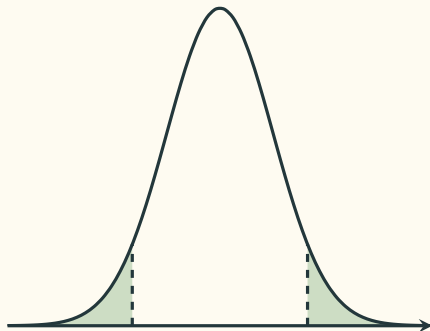
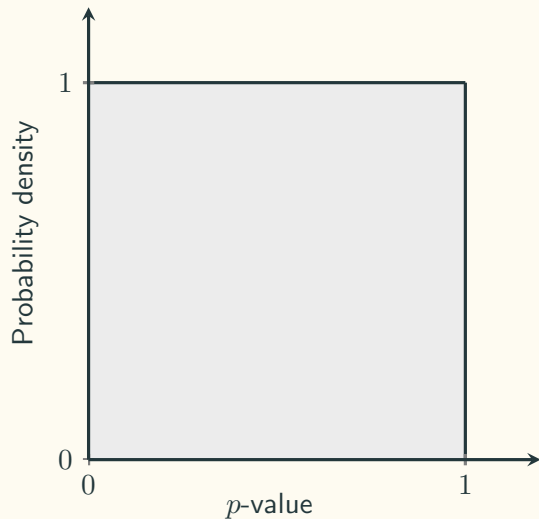
Distribution of p -values When H_0 is true



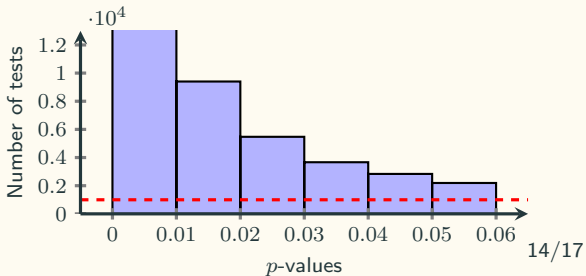
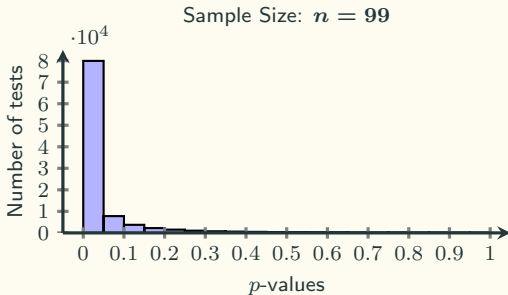
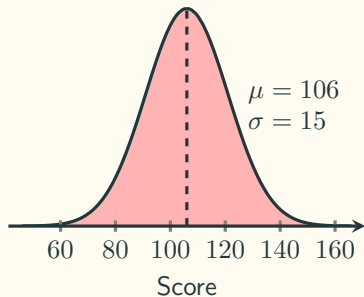
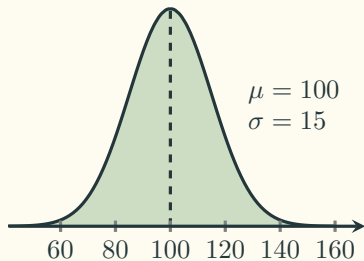
p -value: $\mathbb{P}(\text{data} \mid H_0 \text{ is true})$



p -values Are Uniformly Distributed When H_0 Is True



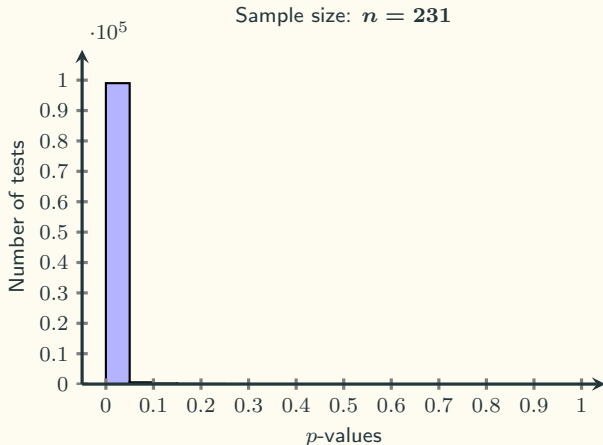
More p -value Distribution When H_1 Is True



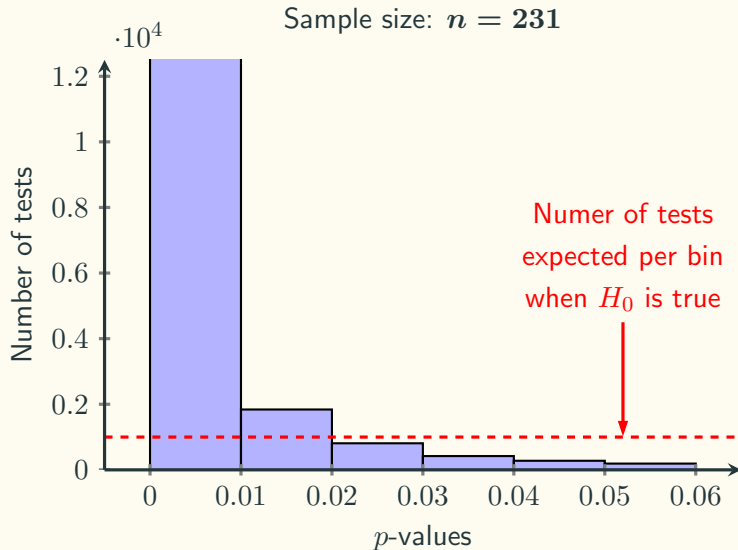
Interpreting p -value When The Power Is High



```
pvals = np.zeros((100000,))  
  
for i in range(100000):  
    s1 = pop1.rvs(size=231)  
    s2 = pop2.rvs(size=231)  
    ts, p = tt(s1, s2, equal_var = True)  
    pvals[i] = p
```



Interpreting p -value When The Power Is High



- **Question:** In this case, if you get a $p = 0.045$ or $p = 0.035$, which one is more likely to be true? H_0 or H_1 ?

Lindley's Paradox (1957)

- In the simulations, we know H_0 is true or not, but in the real world, we don't know. When we have very high power, use an α level of 0.05, and find a p -value of $p = 0.045$, the data is surprising, assuming the null hypothesis H_0 is true, but it is even more surprising, assuming the alternative hypothesis H_1 is true. This shows how a significant p -value is not always evidence for the alternative hypothesis.
- A result can be unlikely when the null hypothesis is true, but it can be even more unlikely assuming the alternative hypothesis is true, and power is very high. For this reason, some researchers have suggested using lower α levels in very large sample sizes, and this is probably sensible advice. Other researchers have suggested using Bayesian statistics, which is also sensible advice.