

# Lecture 30 The Behaviour of The p-value

BIO210 Biostatistics

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Spring, 2022

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南方科技大学生命科学学院  
SUSTech · SCHOOL OF  
**LIFE SCIENCES**

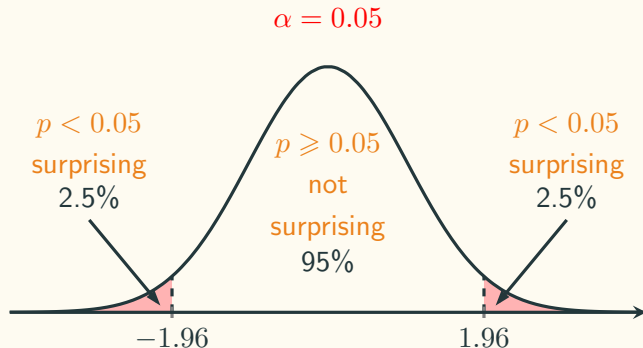
# Why p-values Are Successful In Science

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

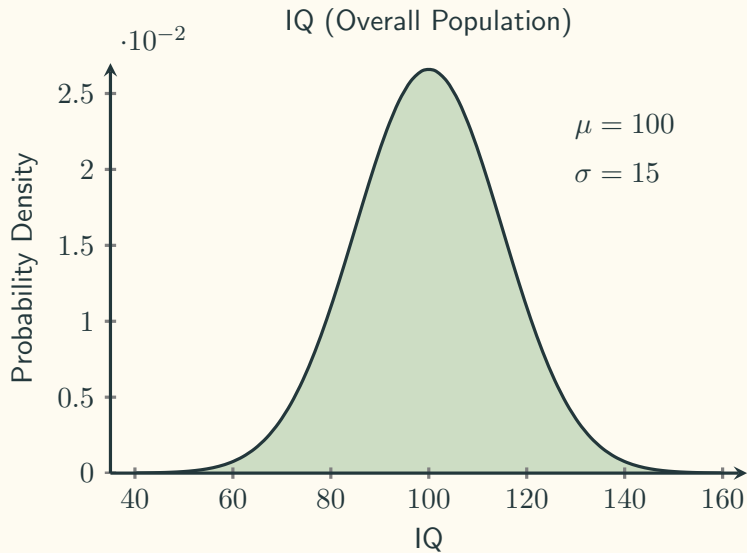
– Benjamini, 2016

# Why p-values Are Successful In Science

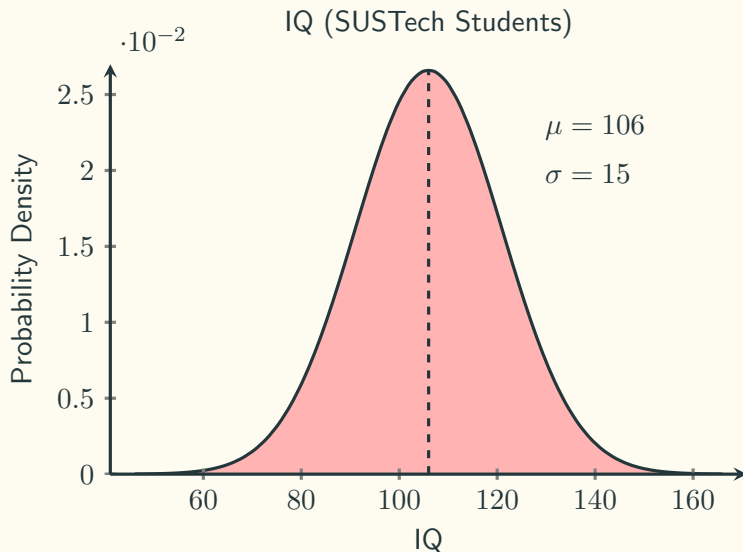
- $p$ -value =  $P(\text{observed data or more extreme} \mid H_0 \text{ is true})$ : How surprising the data is, assuming there is no effect?
- $p$ -value calculation: using the distribution of the test statistics, which is based on the sampling distribution.



## IQ Distribution



# IQ Distribution



Take a sample of  
size  $n = 26$

We ask:

Is  $\mu = 100$  ?

# One Sample $t$ -test

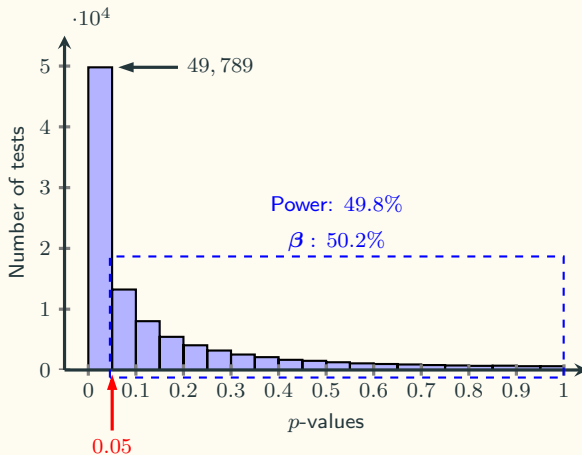


```
import numpy as np
from scipy.stats import ttest_1samp as tt

np.random.seed(42)

pvals = np.zeros((100000,))

for i in range(100000):
    sample = np.random.normal(loc=106,
                              scale=15,
                              size=26)
    ts, p = tt(sample, popmean=100)
    pvals[i] = p
```



# Distribution of $p$ -values

We want to increase our power to 80%:  $n = \left[ \frac{(1.96 + 0.842) \times 15}{106 - 100} \right]^2 = 50$

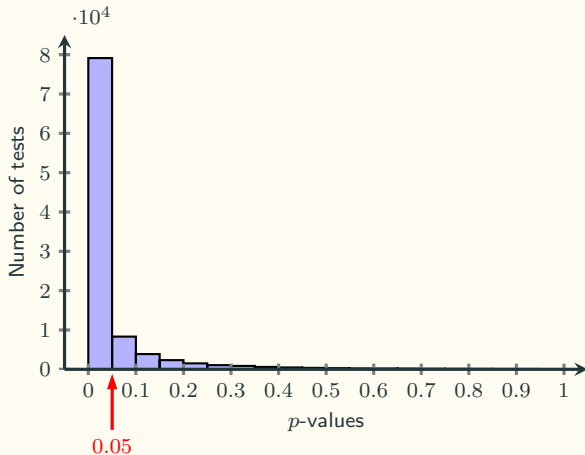


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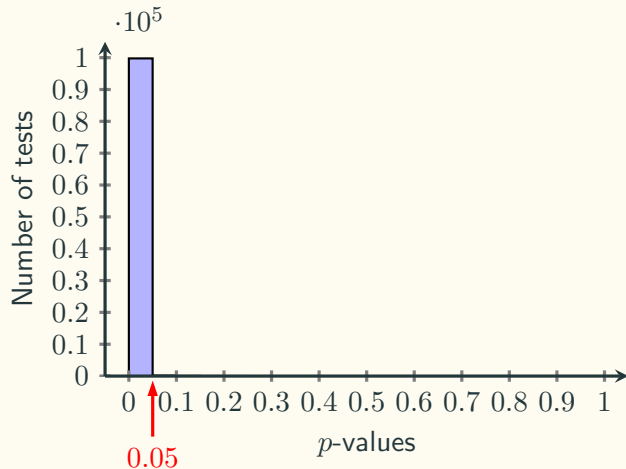
pvals = np.zeros((100000,))

for i in range(100000):
    sample = np.random.normal(loc=106,
                              scale=15,
                              size=50)
    ts, p = tt(sample, popmean=100)
    pvals[i] = p
```



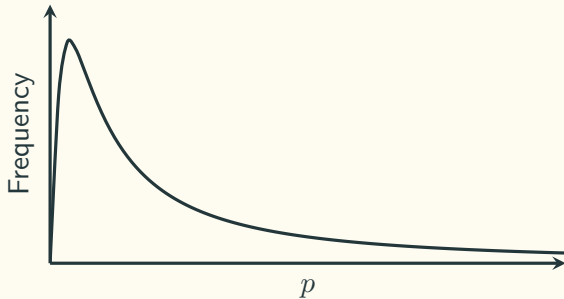
## Distribution of $p$ -values

Sample size:  $n = 144$





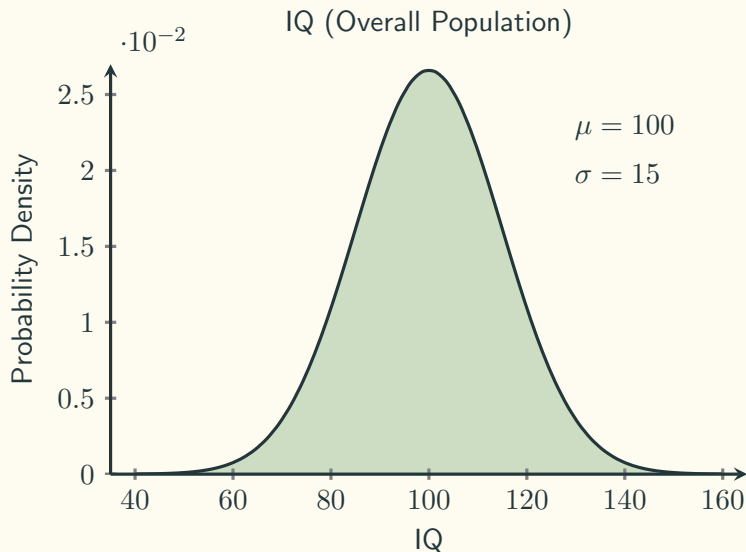
## Distribution of $p$ -values When $H_1$ Is True



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 ?

# IQ Distribution



Take samples of  
size  $n = 100$

We ask:

Is  $\mu = 100$ ?

# $p$ -value Fluctuation

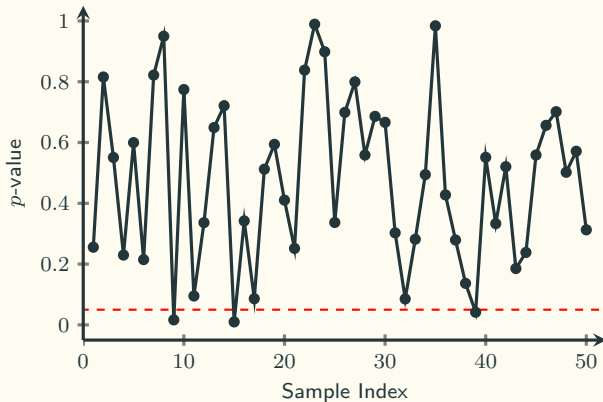


```
import numpy as np
from scipy.stats import ttest_1samp as tt

np.random.seed(42)

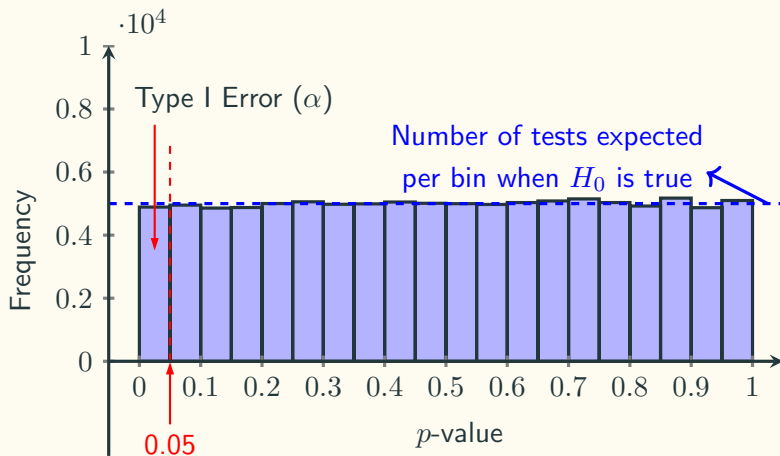
pvals = np.zeros((100000,))

for i in range(100000):
    sample = np.random.normal(loc=100,
                              scale=15,
                              size=100)
    ts, p = tt(sample, popmean=100)
    pvals[i] = p
```



-----  $p = 0.05$

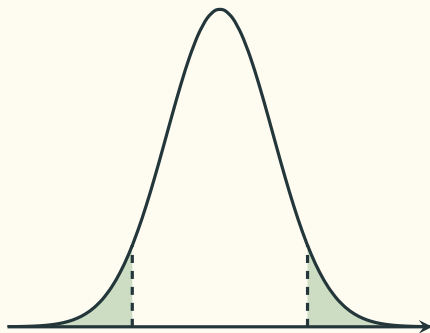
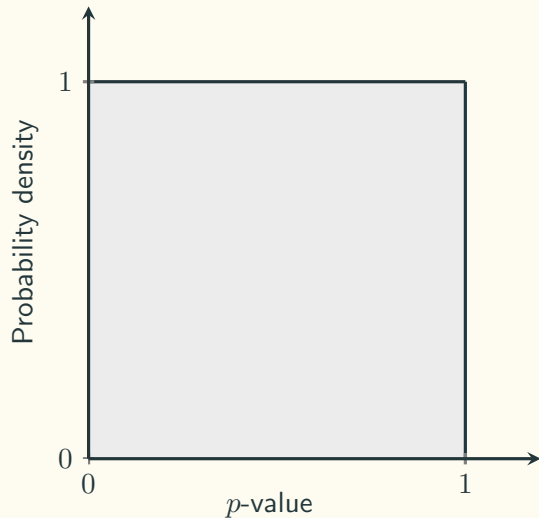
## Distribution of $p$ -values When $H_0$ is true



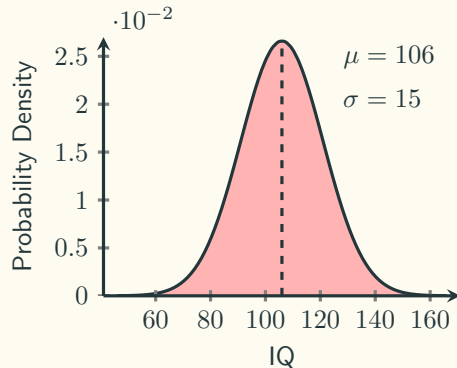
$p$ -value:  $P(\text{data} \mid H_0 \text{ is true})$



## $p$ -values Are Uniformly Distributed When $H_0$ Is True

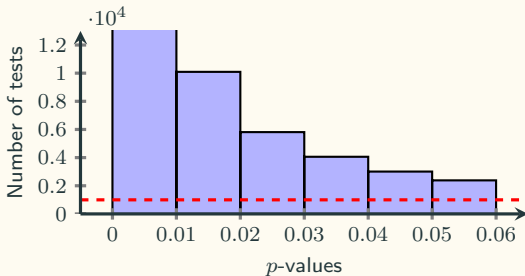
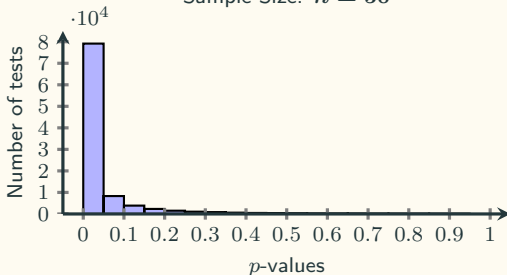


## More $p$ -value Distribution When $H_1$ Is True



Take samples and ask: Is  $\mu = 100$  ?

Sample Size:  $n = 50$



# Interpreting $p$ -value When The Power Is High

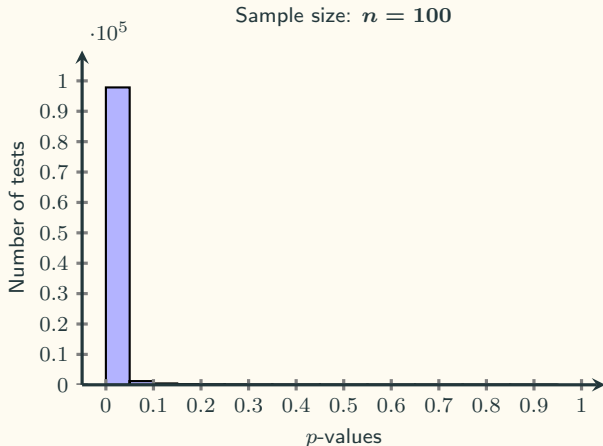


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import numpy as np
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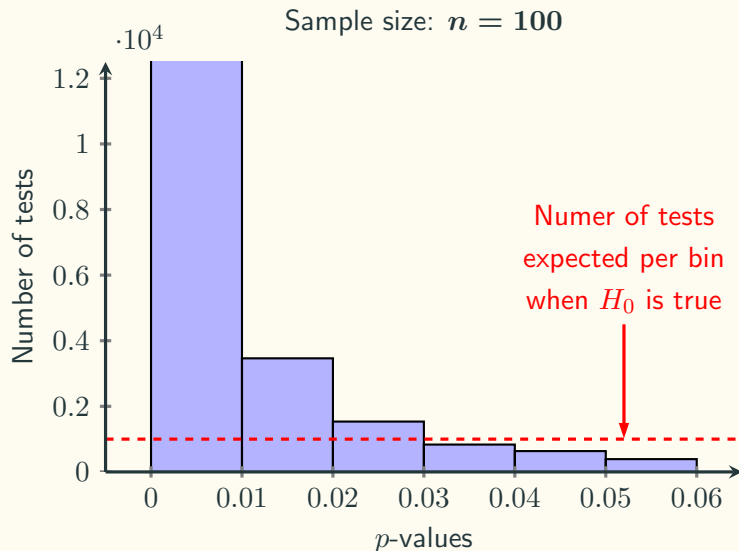
np.random.seed(42)

pvals = np.zeros((100000,))

for i in range(100000):
    sample = np.random.normal(loc=106,
                              scale=15,
                              size=100)
    ts, p = tt(sample, popmean=100)
    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  $\alpha$  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  $\alpha$  levels in very large sample sizes, and this is probably sensible advice. Other researchers have suggested using Bayesian statistics, which is also sensible advice.