05 Two Sample Test

November 20, 2022

1 Two-Sample-Test

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy import stats
```

1.1 Paired data

Let $(X_1, Y_1), \ldots, (X_n, Y_n)$ be paired data such that the Difference $D_i = X_i - Y_i$ satisfies the requirements of the 1-sample t-test. Let $\mu_X = E(X_i)$ and $\mu_Y = E(Y_i)$. Then, $\mu = E(D) = \mu_X - \mu_Y$. Therefore testing

$$H_0: \mu_X = \mu_Y$$
$$H_1: \mu_X \neq \mu_Y$$

can be achieved by performing a 1-sample t-test for

$$H_0: \mu = 0 \text{ vs } H_1: \mu \neq 0$$

Test Statistic

For the test statistic we assume the difference between the two groups.

$$T = \frac{\bar{X} - \bar{Y}}{\frac{s_x}{\sqrt{n}}}$$

```
[2]: x = stats.norm(100, 1).rvs(100)
y = x - stats.norm(1, 5).rvs(100)

d = x - y

# Perform the test
alpha = 0.05
d_mean = np.mean(d)
std = np.std(d, ddof=1)
```

```
c = stats.t(d.shape[0] - 1).ppf(1 - alpha / 2) # right-sided

t = (d_mean) / (std / np.sqrt(d.shape[0]))

p_value = 1 - stats.norm().cdf(t)

print(f"Rejection area: [{c}; inf]")
print(f"T-statistics: {t}")
print(f"P-Value: {p_value}")
print(f"Is significant: {p_value < alpha}")</pre>
```

Rejection area: [1.9842169515086827; inf]

T-statistics: 1.006216961461008 P-Value: 0.15715560739037338

Is significant: False

1.2 Differences in two groups

We assume that X and Y are independent random variables with

$$X \sim N(\mu, \sigma_X^2)$$

 $Y \sim N(\mu, \sigma_Y^2)$

for a given $\Delta \in \mathbb{R}$

$$H_0: \mu_X - \mu_Y = \Delta$$

against the alternative

$$H_1: \mu_x - \mu_y \leq \Delta$$
 or
 $H_1: \mu_x - \mu_y \neq \Delta$ or
 $H_1: \mu_x - \mu_y \geq \Delta$

1.2.1 Equal variance

Let X_1, \ldots, X_n and Y_1, \ldots, Y_n be independent samples of X and Y with

$$\sigma_x = \sigma_Y = \sigma$$

Given H_0 is true, we find that the statistic

$$T = \frac{\bar{X} - \bar{Y} - \Delta}{s_P * \sqrt{\frac{1}{n} + \frac{1}{m}}}$$

follows a student-t distribution with m+n-2 degrees of freedom. Here

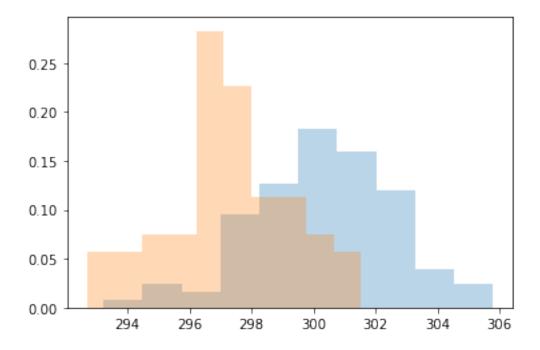
$$s_P^2 = \frac{(n-1)s_X^2 + (m-1)s_Y^2}{n+m-2}$$

is the pooled variance

```
[5]: x = stats.norm(300, 2).rvs(100)
     y = stats.norm(297, 2).rvs(60)
     delta = 3 # Assumed difference in the group
     x_mean = np.mean(x)
     y_{mean} = np.mean(y)
     x_var = np.var(x, ddof=1)
     y_var = np.var(y, ddof=1)
     n = x.shape[0]
     m = y.shape[0]
     c = stats.t(n + m - 2).ppf(1 - 0.05 / 2)
     sp = np.sqrt(((n - 1) * x_var + (m - 1) * y_var) / (n + m - 2))
     t = (x_mean - y_mean - delta) / (sp * np.sqrt(1 / n + 1 / m))
     p = 1 - stats.t(n + m - 2).cdf(t)
     print(f"T-Statistic: {t}")
     print(f"Rejection a: {c}")
     print(f"P-Value:
                          {p}")
    T-Statistic: 0.18942639773551548
    Rejection a: 1.975092072704601
    P-Value:
                 0.42500088002587844
[6]: t, p = stats.ttest_ind(x, y + delta, equal_var=True, alternative='greater')
     print(f"T-Statistic: {t}")
     print(f"P-Value:
```

T-Statistic: 0.18942639773551548 P-Value: 0.42500088002587844

```
[7]: plt.hist(x, density=True, alpha=.3)
plt.hist(y, density=True, alpha=.3)
plt.show()
```



1.2.2 Welch Test

- If the variance of X and Y are different, then the pooled variance S_P can not be computed
- The test statistic T is computed similarly, but the distribution under H_0 is now known
- \bullet It is possible to approximate the distribution of T by a t-distribution
- In this case the test is called the Welch-Test

```
[8]: x = stats.norm(300, np.random.randint(1, 3)).rvs(1000)
y = stats.norm(297, np.random.randint(1, 3)).rvs(600)
delta = 3

n = x.shape[0]
m = y.shape[0]
alpha = 0.05

c = stats.t(n + m - 2).ppf(1 - alpha / 2)
```

```
[9]: t, p = stats.ttest_ind(x, y + delta, equal_var=False, alternative='greater')

print(f"T-Statistic: {t}")
print(f"Rejection a: {c}")
print(f"P-Value: {p}")
```

T-Statistic: 0.2532131916753035 Rejection a: 1.9614496156420809 P-Value: 0.40007234294834687 []:[