# **Appendix**

In what follows, we have provided the mathematical explanation of our work.

### Statistical Model

Let  $P = \{P_{\theta} : \theta \in \Theta\}$  be a parametrized statistical model and the random variable X represent the suicide rates i.e., the number of suicides per one hundred thousand of the population. We assume that X admits the density  $P_{\theta}$ .

The dataset includes suicide rates from 1985 until 2016 which have 27820 rows and 7 columns. The column 'suicide rates' forms our observations  $X_i$ .

Since we are intent on investigating the role of gender in our dataset, we split the dataset based on the *gender* column and formulate two sets of observations for males and females. The samples  $X_1^M, ..., X_{n_1}^M$  and  $X_1^F, ..., X_{n_2}^F$  represent males and females' suicide rates, respectively.

#### **Parameters**

We denote the mean of the suicide rates for population of males and females by  $\mu_M$  and  $\mu_F$  respectively.  $n_1$  and  $n_2$  are the size of the samples of males and females, respectively and both are equal to 13910.

## Hypothesis Test

We aim at testing the hypothesis that the mean of population for men is less than mean of population for women. For this purpose, we investigate the quantity  $\theta = \mu_M - \mu_F$ . For testing our hypothesis, we define our test as following:

$$H_0: \mu_M < \mu_F$$
 against  $H_1: \mu_M > \mu_F$ 

Using notation of  $\theta$ , the test can also be written as

$$H_0: \theta \leq 0$$
 against  $H_1: \theta > 0$ 

Consequently, we have  $\Theta_0 = \{\theta : \theta \leq 0 \}, \ \Theta_1 = \{\theta : \theta > 0 \}$ 

#### Remarks

Since  $\sigma$  for each sample is unknown, we estimate it using the sample variance of two samples which are defined as following

$$S_M^2 = \frac{1}{n_1 - 1} \sum_{i=1}^{n_1} (X_i^M - \overline{X}^M)^2$$
  $S_F^2 = \frac{1}{n_2 - 1} \sum_{i=1}^{n_2} (X_i^F - \overline{X}^F)^2$ 

And the total sample variance is

$$S_T^2 = \frac{S_M^2}{n_1} + \frac{S_F^2}{n_2}$$

Using central limit theorem, we can show that if  $n \to \infty$ , then

$$\overline{X}^M - \overline{X}^F \xrightarrow{L} \mathcal{N}(\mu_M - \mu_F, \sigma_M^2 + \sigma_F^2)$$
.

Therefore, by the definition of student t's distribution, we have

$$\frac{\left(\overline{X}^M - \overline{X}^F - (\mu_M - \mu_F)\right)}{S_T} \xrightarrow{\mathcal{L}} t_{df = n_1 + n_2 - 2}$$

#### **Estimator**

As we are interested in the value  $\theta = \mu_M - \mu_F$ , we define an estimator of it by

$$\overline{X}^M - \overline{X}^F$$

We claim that this is a consistent estimator of  $\theta$ . To prove consistency, we can say that by law of large numbers, if n is large enough, then

$$\overline{X}^M - \overline{X}^F \xrightarrow{a.s.} \mathbb{E}[X_1^M] - \mathbb{E}[X_1^F] = \mu_M - \mu_F$$

#### Test Statistic

In order to accept or reject the test, we define a test statistic as following

$$T(X) = \overline{X}^M - \overline{X}^F$$

#### Intuitive Approach: finding the threshold

We will use the following quantity as our pivot.

$$\tau = \frac{\overline{X}^M - \overline{X}^F - (\mu_M - \mu_F)}{S_T} \sim t_{df = n_1 + n_2 - 2}$$

where degrees of freedom is  $n_1 + n_2 - 2$ . For  $\theta \in \Theta_0$  or equivalently  $\theta \leq 0$ ,

$$\alpha = \mathbb{P}(T(X) > u)$$

$$= \mathbb{P}(\overline{X}^M - \overline{X}^F > u)$$

$$= \mathbb{P}(\frac{\overline{X}^M - \overline{X}^F - (\mu_M - \mu_F)}{S_T} > \frac{u - (\mu_M - \mu_F)}{S_T})$$

$$= \mathbb{P}(\frac{\overline{X}^M - \overline{X}^F - \theta}{S_T} > \frac{u - \theta}{S_T})$$

$$< \mathbb{P}(\frac{\overline{X}^M - \overline{X}^F - \theta}{S_T} > \frac{u}{S_T})$$

$$= \mathbb{P}(\tau > \frac{u}{S_T})$$

$$= \mathbb{P}(\tau > t_{\alpha})$$

$$(1)$$

where  $t_{\alpha} = \frac{u}{S_T}$  is the  $\{1 - \alpha\}$  quantile of the t-distribution and  $\tau$  is the pivot. The last inequality comes from the fact that since  $\theta < 0$ , we have  $u - \theta > u$ . Solving for threshold u gives that  $u = S_T t_{\alpha}$ .

Then the test function  $\phi$  will be

$$\phi(X) = \mathbb{1}_{\{T(X)>u\}} \tag{2}$$

Substituting u into test function will give us the decision

$$\phi(X) = \mathbb{1}_{\{T(X) > S_T t_\alpha\}} \tag{3}$$

#### Power of Test

The power of test is expressed in the following

$$\mathbb{P}(rejectH_{0}) = \mathbb{P}(\Phi(X) = 1) = \mathbb{E}[\Phi] 
= \mathbb{P}(\overline{X}^{M} - \overline{X}^{F} > u) 
= \mathbb{P}(\frac{\overline{X}^{M} - \overline{X}^{F} - (\mu_{M} - \mu_{F})}{S_{T}} > \frac{u - (\mu_{M} - \mu_{F})}{S_{T}}) 
= \mathbb{P}(\frac{\overline{X}^{M} - \overline{X}^{F} - \theta}{S_{T}} > \frac{u - \theta}{S_{T}}) 
= 1 - \phi(\frac{u - \theta}{S_{T}})$$
(4)

where  $\phi$  is the cdf of the t-distribution.

For  $\theta \in \Theta_1$  or equivalently  $\theta > 0$ , since  $\theta > 0$ ,  $u - \theta$  is a decreasing function of  $\theta$ . As  $\phi$  is an increasing function,  $\phi(\frac{u-\theta}{S_T})$  is also decreasing. Therefore,  $1 - \phi(\frac{u-\theta}{S_T})$  will be a increasing function of  $\theta$ .

#### Results

$$S_M = 23.55,$$
  $S_F = 7.35,$   $S_T = 0.21,$   $T(X) = 14.85,$   $t_{alpha} = 1.65,$   $S_T = 0.21,$   $u = 0.34$ 

Since T(x) > u, we reject the null hypothesis and we increase our confidence in  $H_1$ , which states that the mean for males is greater that mean for females.