



Homework #3

Due by Sunday 11/22 23:55

Question #1.A

Sudden infant death syndrome (SIDS) is the sudden and unexplained death of a baby younger than 1 year old. A diagnosis of SIDS is made if the baby's death remains unexplained even after a death scene investigation, an autopsy, and a review of the clinical history. The low birth weight children are at risk, and these question compare the distribution between boys and girls.

Question #1.B

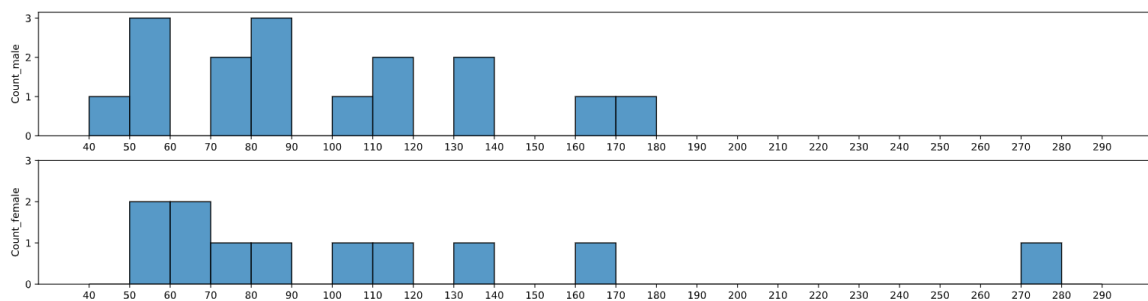


Figure 1: Histogram

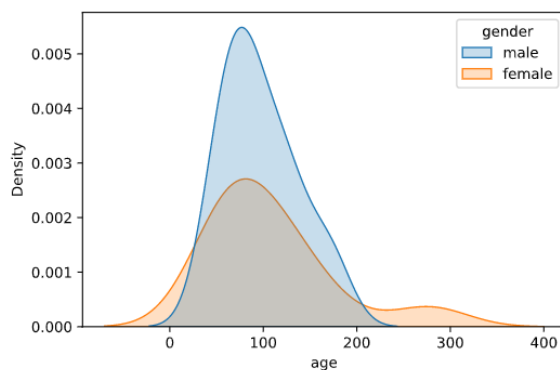


Figure 2: KDE Plot

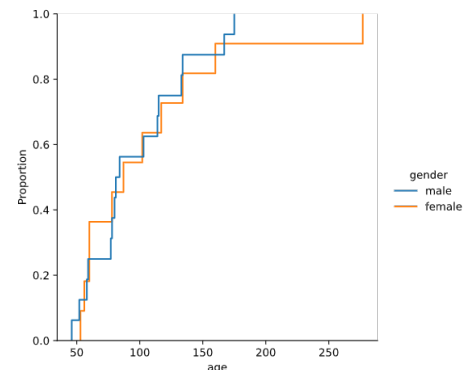


Figure 3: Empirical distribution

成對樣本T檢定的case通常有兩種，一種是「重複量數」，也就是前後測的問題，像是減肥前後、實驗前後的成績，因此每一筆配對的資料都是來自同一位受試者，這種是最常見的配對樣本。但這筆資料不是同一個人，意思是不是從男生變女生。

另一種為「配對組法」，雖然每一筆配對的資料是來自兩位的受試者，但是我們會認定他們的某一特質（研究者所關心的）是相同的。意思是，想擁有一些手法，讓那配對其他條件都相同下，只有男女這一個差別，那這樣才適合做T 檢定。

也就是說，不能直接拿男女的存活時間直接做T檢定，即便他們是抽樣出來的。

Question #1.C

The result from `scipy.stats.ttest_ind` is:

```
statistic = 0.5118640907091607  
pvalue = 0.6132382438368118
```

Since we have $pvalue = 0.61324 > 0.05$, that is, we cannot reject the null hypothesis at confidence interval with level 95%. Therefore, the two distributions are the same.

Question #1.D

The result from `scipy.stats.ranksums` is:

```
statistic = 0.148039131365948  
pvalue = 0.8823118861000214
```

Since we have $pvalue = 0.8823 > 0.05$, that is, we cannot reject the null hypothesis at confidence interval with level 95%. Therefore, the two distributions are the same.

Also, we get a higher p-value than t-test makes sense, since Wilcoxon rank-sum test is more appropriate than t-test for the question.

Question #1.E

To answer the question, we pool the two samples, order the observations from the smallest to the largest, retaining their group identity, and rank them from 1 to 27.

Since the sample of female ($m=11$) is smaller than that of male ($n=16$), the test statistic W_m is the sum of the ranks associated with the female.

Therefore, $W_m = 3 + 4 + 7.5 + \dots + 27 = 157$

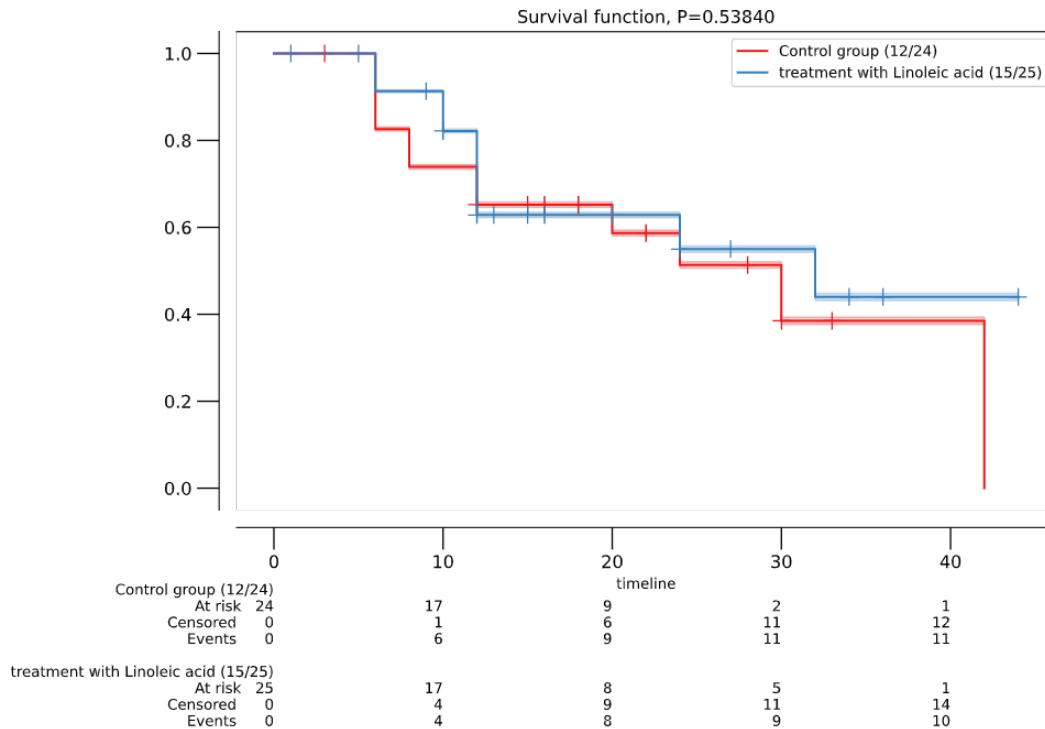
And, $E(W) = \frac{11(11+16+1)}{2} = 154$,

$Var(W) = \sqrt{\frac{11 \cdot 16(11+16+1)}{12}} \rightarrow SD = 20.26$

$Z = \frac{157-154}{20.26} = 0.148$

The p-value with z-score=0.148 is .882343 under two-tail hypothesis. The result is very close to (d), and also we cannot reject $H_0 : F_1 = F_2$.

Question #2.A



The horizontal axis (x-axis) represents time in days, and the vertical axis (y-axis) shows the probability of surviving or the proportion of people surviving. The lines represent survival curves of the two groups. A vertical drop in the curves indicates an event. The vertical tick mark on the curves means that a patient was censored at this time.

- At time zero, the survival probability is 1.0 (or 100% of the participants are alive).
- The median survival is approximately 30 days for Control group and 32 days for treatment with Linoleic acid, suggesting a good survival for treatment with Linoleic acid compared to Control group

Question #2.B

The result from `lifelines.statistics.logrank_test` is:

statistic = 0.378521
pvalue = 0.538396

The result from `gehan.test` is:

statistic = 0.3345791
pvalue = 0.5629751

The log rank test for difference in survival gives a p-value of $p = 0.538396$ and the Gehan test gives $p = 0.5629751$, indicating that there is no evidence that treatment with Linoleic acid have better effect in survival at confidence interval with level 95%. Also, the fact that we get a larger p-value from Gehan test compared to the log-rank test makes sense

Question #3.A

$$l(\lambda) = \log L(\lambda) = \sum_{i=1}^n \delta_i (\log \lambda - \lambda x_i) + (1 - \delta_i)(-\lambda x_i) \quad (1)$$

To find MLE of λ , let $\frac{\partial l(\lambda)}{\partial \lambda} = 0$, that is:

$$\frac{\partial l(\lambda)}{\partial \lambda} = \sum_{i=1}^n \frac{\delta_i}{\lambda} - x_i = 0 \quad (2)$$

$$\lambda_{MLE} = \frac{\sum_{i=1}^n \delta_i}{\sum_{i=1}^n x_i} \quad (3)$$

Question #3.C

$$\frac{\partial^2 l(\lambda)}{\partial \lambda^2} = \frac{\partial}{\partial \lambda} \sum_{i=1}^n \frac{\delta_i}{\lambda} - x_i = -\frac{\sum_{i=1}^n \delta_i}{\lambda^2} \quad (4)$$

$$Var(\hat{\lambda}) = -\left(\frac{\sum_{i=1}^n \delta_i}{\lambda^2}\right)^{-1} = \frac{\sum_{i=1}^n \delta_i}{\left(\sum_{i=1}^n x_i\right)^2} \quad (5)$$

Question #4.A

$$\begin{aligned}
P(Y = y) &= \binom{y+r-1}{y} p^y (1-p)^r \\
&= \exp(y \ln(p) + r \ln(1-p) + \ln(\binom{y+r-1}{y}))
\end{aligned} \tag{6}$$

Thus the canonical parameter $\theta = \ln(p)$, and the canonical link function $g(\mu) = \ln(\frac{\mu/r}{1+\mu/r})$

Question #4.B

Continue from A, we have canonical parameter $\theta = \ln(p)$, that is, $p = e^\theta$

$$P(Y = y) = \exp(\theta y + r \ln(1 - e^\theta) + \ln(\binom{y+r-1}{y})) \tag{7}$$

That is, we have $b(\theta) = -r \ln(1 - e^\theta)$ and $\phi = 1$.

$$E[y] = \frac{\partial b(\theta)}{\partial \theta} = r \frac{e^\theta}{1 - e^\theta} = r \frac{p}{1 - p} = \frac{rp}{1 - p} \tag{8}$$

$$Var[y] = \phi \frac{\partial^2 b(\theta)}{\partial \theta^2} = \frac{\partial}{\partial \theta} r \frac{e^\theta}{1 - e^\theta} = \frac{r e^\theta}{(1 - e^\theta)^2} = \frac{rp}{(1 - p)^2} \tag{9}$$