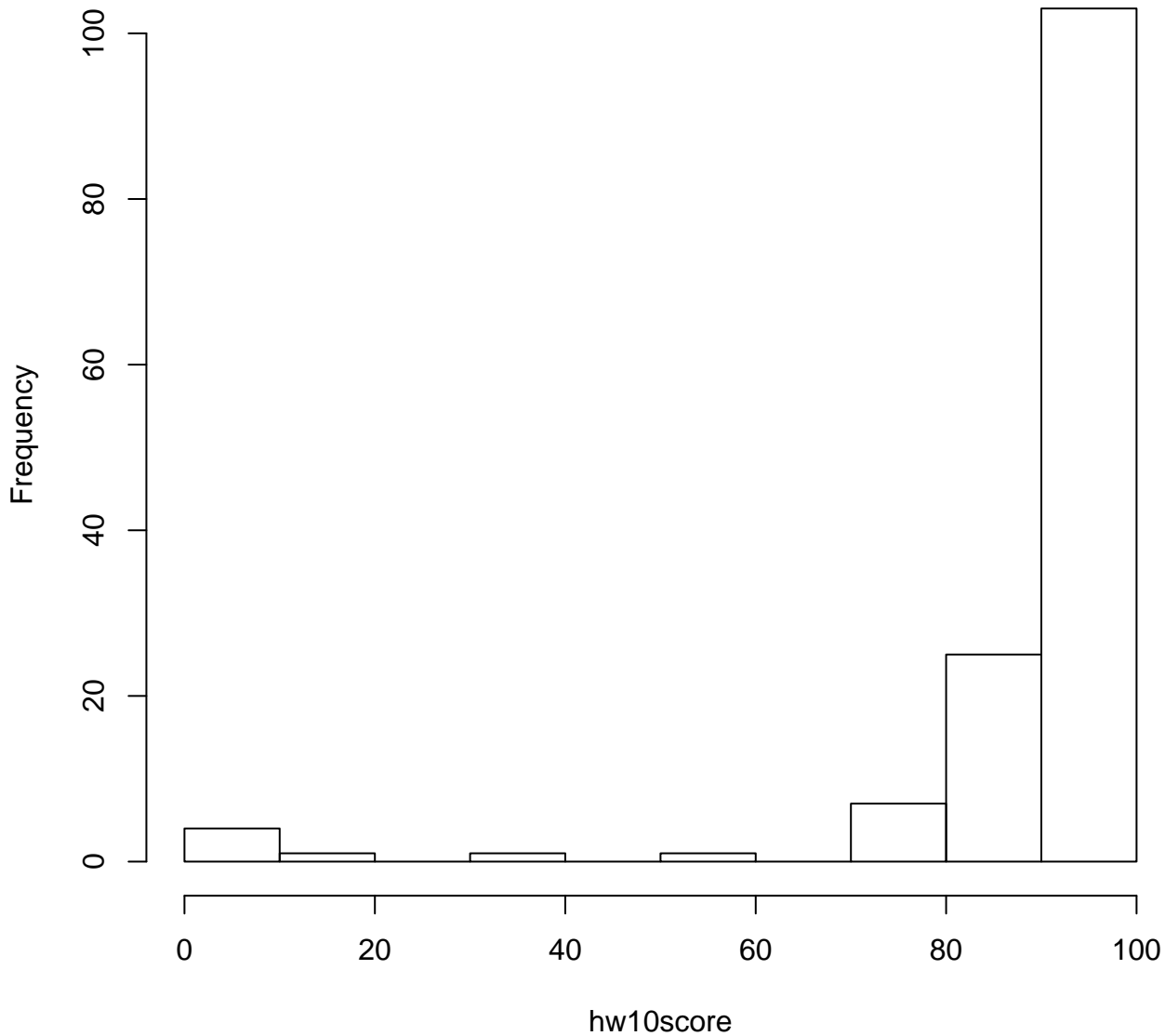


Histogram of hw10score



$$\alpha = 0.05$$

$$4. \quad \bar{x}_{..} = \frac{1}{n_T} \sum_{i=1}^k n_i \cdot \bar{x}_{i.}$$

$$SSTr = \sum_{i=1}^k n_i (\bar{x}_{i.} - \bar{x}_{..})^2 \quad (+10)$$

$$SSE = SST - SSTr.$$

Using the above, ANOVA Table is :

	df	SS	MS	F	P
Tr	6	7.6138	1.269	0.7782	0.5895.
E	77	125.462	1.6307		
Total	83	133.18			

(+10)

$$8. \quad n_T = 33, \quad k = 3, \quad f_{0.05, 3, 30} = 3.49, \quad MSE = 4.96.$$

Formula for 95% CI of $\mu_i - \mu_j$ is

$$\bar{x}_{i.} - \bar{x}_{j.} \pm \sqrt{MSE} \frac{f_{0.05, 3, 30}}{\sqrt{2}} \sqrt{\frac{1}{n_i} + \frac{1}{n_j}}.$$

$$a) \quad 1, 2 : (0.9665, 5.6535)$$

$$1, 3 : (-3.4035, 1.2835) \ni 0 \quad \leftarrow 1, 3 : \text{not distinguishable.}$$

$$2, 3 : (-6.7135, -2.0265)$$

(+8)

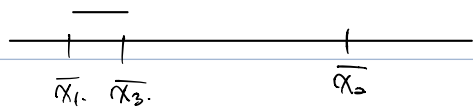
$$b) \quad \begin{array}{c} \text{---} \\ | \quad | \quad | \\ \bar{x}_{2.} \quad \bar{x}_{1.} \quad \bar{x}_{3.} \end{array} \quad (+6)$$

$$c) \quad L^2 = \frac{4MSE f_{0.05, 3, 30}^2}{n} \leq 4 \quad \Rightarrow \quad n \geq MSE \cdot f_{0.05, 3, 30}^2 = 60.xx$$

$$\therefore n \geq 61, \quad \text{so } 50 \text{ more.} \quad (+6)$$

16. See the Table below.

$p < 0.05$. (reject the null)



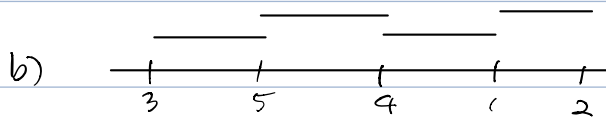
\Rightarrow There is an evidence that

layout 2 is the slowest. (+5)

ANOVA Table (+5), pairwise CI (+10).

20. a) ANOVA Table (+10)

$p < 0.05$.

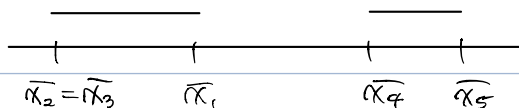


$\therefore (3,5)$: smallest, $(1,2)$: largest.

by observing pairwise CI (+10)

24. ANOVA Table (+10)

$p < 0.05$.



\therefore smallest : $(1,2,3)$, largest : $(4,5)$.

by observing pairwise CI (+10)

Importing packages, modules and data

```
In [1]: import pandas as pd
df1 = pd.read_excel('data/ds11.1.4-keyboard-layout-designs.xls')
df2 = pd.read_excel('data/ds11.1.7-oneway-layout-3.xls')
df3 = pd.read_excel('data/ds11.1.9-ecoli-colonies-in-riverwater.xls')
```

```
In [2]: import scipy.stats as stat
from statsmodels.stats.anova import anova_lm
from statsmodels.formula.api import ols
from statsmodels.stats.multicomp import pairwise_tukeyhsd
```

ANOVA Table for 16

```
In [3]: aq = df1
newaq = pd.melt(aq).dropna()
model = ols('value ~ C(variable)', newaq).fit()
print(anova_lm(model))
```

	df	sum_sq	mean_sq	F	PR(>F)
C(variable)	2.0	121.238197	60.619098	52.83944	1.476166e-10
Residual	30.0	34.416961	1.147232	NaN	NaN

CI for 16 ¶

```
In [4]: comp = pairwise_tukeyhsd(newaq['value'], newaq['variable'], alpha=0.05)
print(comp)
```

```
Multiple Comparison of Means - Tukey HSD, FWER=0.05
=====
group1  group2  meandiff p-adj  lower  upper  reject
-----
Layout 1 Layout 2    3.985  0.001  2.8546  5.1154   True
Layout 1 Layout 3   -0.3614  0.6879 -1.4633  0.7406  False
Layout 2 Layout 3   -4.3464  0.001 -5.4998 -3.1929   True
=====
```

ANOVA Table for 20

```
In [5]: aq = df2
newaq = pd.melt(aq).dropna()
model = ols('value ~ C(variable)', newaq).fit()
print(anova_lm(model))
```

	df	sum_sq	mean_sq	F	PR(>F)
C(variable)	4.0	1102.74	275.685	18.50856	0.000002
Residual	20.0	297.90	14.895	NaN	NaN

CI for 20

```
In [6]: comp = pairwise_tukeyhsd(newaq['value'], newaq['variable'], alpha=0.05)
print(comp)
```

```
Multiple Comparison of Means - Tukey HSD, FWER=0.05
=====
  group1    group2   meandiff p-adj   lower   upper   reject
-----
Treatment 1 Treatment 2      6.4 0.1039  -0.9046  13.7046   False
Treatment 1 Treatment 3     -13.2 0.001  -20.5046  -5.8954    True
Treatment 1 Treatment 4      -2.6 0.834  -10.3477   5.1477   False
Treatment 1 Treatment 5      -7.1 0.0455  -14.0937  -0.1063    True
Treatment 2 Treatment 3     -19.6 0.001  -26.9046 -12.2954    True
Treatment 2 Treatment 4      -9.0 0.018  -16.7477  -1.2523    True
Treatment 2 Treatment 5     -13.5 0.001  -20.4937  -6.5063    True
Treatment 3 Treatment 4      10.6 0.0046   2.8523  18.3477    True
Treatment 3 Treatment 5       6.1 0.1063  -0.8937  13.0937   False
Treatment 4 Treatment 5      -4.5 0.3988 -11.9553   2.9553   False
=====
```

ANOVA Table for 24

```
In [7]: aq = df3
newaq = pd.melt(aq).dropna()
model = ols('value ~ C(variable)', newaq).fit()
print(anova_lm(model))
```

	df	sum_sq	mean_sq	F	PR(>F)
C(variable)	4.0	596.3	149.075	24.438525	0.000002
Residual	15.0	91.5	6.100	NaN	NaN

CI for 24

```
In [8]: comp = pairwise_tukeyhsd(newaq['value'], newaq['variable'], alpha=0.05)
print(comp)
```

Multiple Comparison of Means - Tukey HSD, FWER=0.05

group1	group2	meandiff	p-adj	lower	upper	reject
Location 1	Location 2	-0.25	0.9	-5.6435	5.1435	False
Location 1	Location 3	-0.25	0.9	-5.6435	5.1435	False
Location 1	Location 4	8.0	0.0028	2.6065	13.3935	True
Location 1	Location 5	13.0	0.001	7.6065	18.3935	True
Location 2	Location 3	0.0	0.9	-5.3935	5.3935	False
Location 2	Location 4	8.25	0.0022	2.8565	13.6435	True
Location 2	Location 5	13.25	0.001	7.8565	18.6435	True
Location 3	Location 4	8.25	0.0022	2.8565	13.6435	True
Location 3	Location 5	13.25	0.001	7.8565	18.6435	True
Location 4	Location 5	5.0	0.0757	-0.3935	10.3935	False