

Booklet of Code and Output  
for  
STAC32 Midterm Exam

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```
library(tidyverse)

## -- Attaching packages -----
tidyverse 1.2.1 --
## v ggplot2 3.2.1    v purrr 0.3.2
## v tibble 2.1.3    v dplyr 0.8.3
## v tidyr 1.0.0     v stringr 1.4.0
## v readr 1.3.1    v forcats 0.4.0
## -- Conflicts -----
tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

library(smmr)
library(PMCMRplus)
```

Figure 1: Packages

height	frequency	vegetation
0.05m	1-per-year	17.3
0.05m	1-per-year	19.3
0.05m	1-per-year	15
0.05m	1-per-year	16.7
0.10m	1-per-year	16
0.10m	1-per-year	15.6
0.10m	1-per-year	16.9
0.10m	1-per-year	15
0.20m	1-per-year	16.7
0.20m	1-per-year	17.9
0.20m	1-per-year	15.9
0.20m	1-per-year	13.7
0.05m	2-per-year	22.4
0.05m	2-per-year	20.8
0.05m	2-per-year	24.5
0.05m	2-per-year	21.7
0.10m	2-per-year	23.9
0.10m	2-per-year	23.6
0.10m	2-per-year	21.7
0.10m	2-per-year	23.8
0.20m	2-per-year	24.7
0.20m	2-per-year	26.3
0.20m	2-per-year	27.2
0.20m	2-per-year	26.4
0.05m	3-per-year	18.6
0.05m	3-per-year	17.9
0.05m	3-per-year	16.1
0.05m	3-per-year	19.4
0.10m	3-per-year	22.2
0.10m	3-per-year	25.6
0.10m	3-per-year	21.8
0.10m	3-per-year	23.6
0.20m	3-per-year	27
0.20m	3-per-year	25.3
0.20m	3-per-year	23.8
0.20m	3-per-year	28

Figure 2: Texas highway mowing data

uptake  
7251  
6871  
9632  
6866  
9094  
5849  
8957  
7978  
7064  
7494  
7883  
8178  
7523  
8724  
7468

Figure 3: Uptake rate data

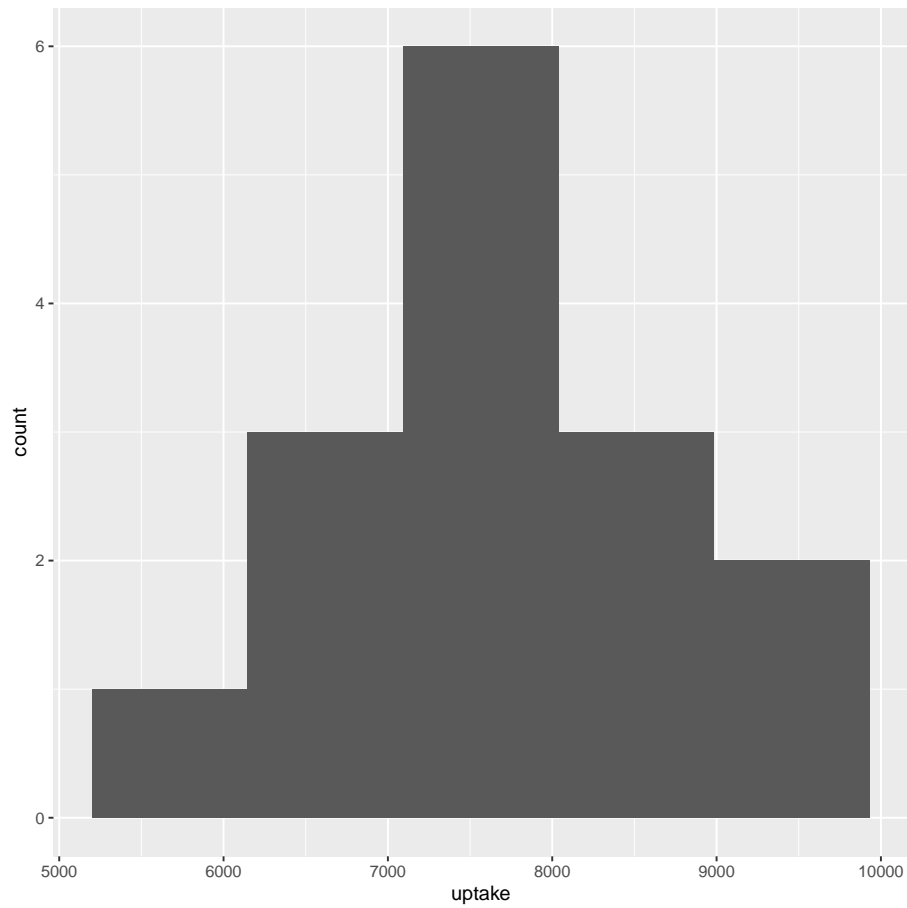


Figure 4: Histogram of uptake rates

```
##
## One Sample t-test
##
## data: uptake
## t = -0.81599, df = 14, p-value = 0.2141
## alternative hypothesis: true mean is less than 8000
## 95 percent confidence interval:
##      -Inf 8244.674
## sample estimates:
## mean of x
##      7788.8
```

Figure 5:  $T$ -test output

```
##
## One Sample t-test
##
## data: uptake
## t = 30.093, df = 14, p-value = 3.998e-14
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 7233.672 8343.928
## sample estimates:
## mean of x
## 7788.8
```

Figure 6: 95% confidence interval for population mean uptake rate

```
## # A tibble: 2 x 2
##   `pvals < 0.05`      n
##   <lgl>              <int>
## 1 FALSE              587
## 2 TRUE               413
```

Figure 7: Output from power estimation analysis

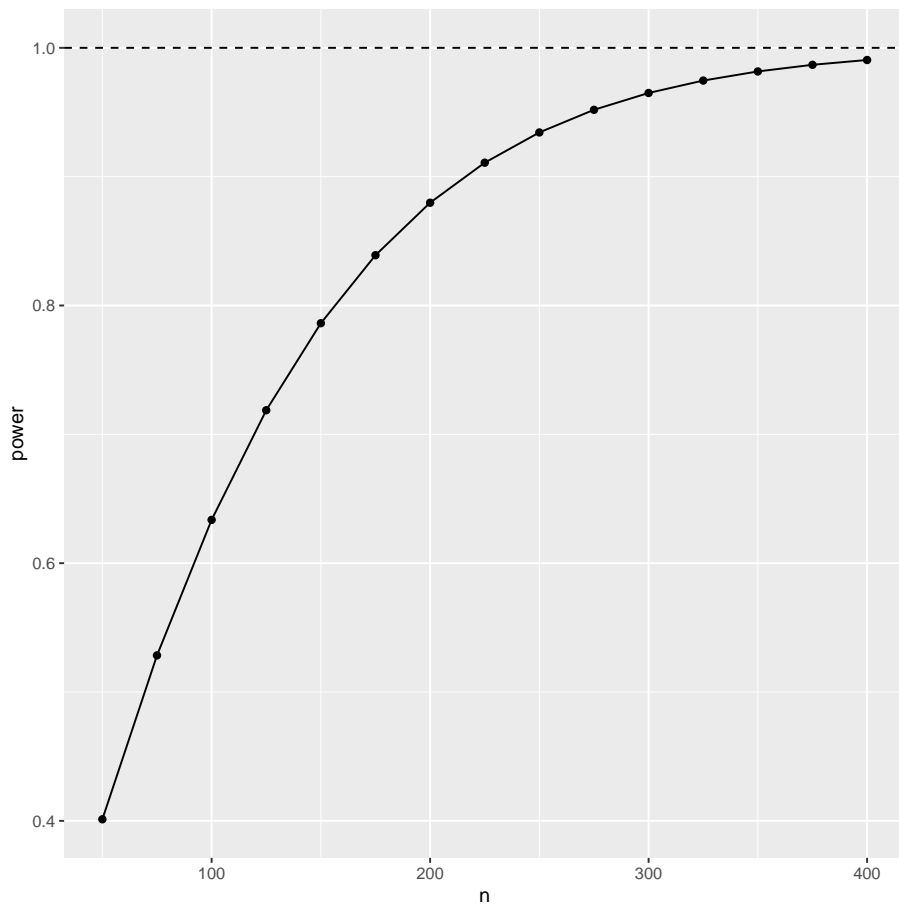


Figure 8: Power curve for different sample sizes



```
ggplot(wrist, aes(x=extension)) + geom_histogram(bins=10)
```

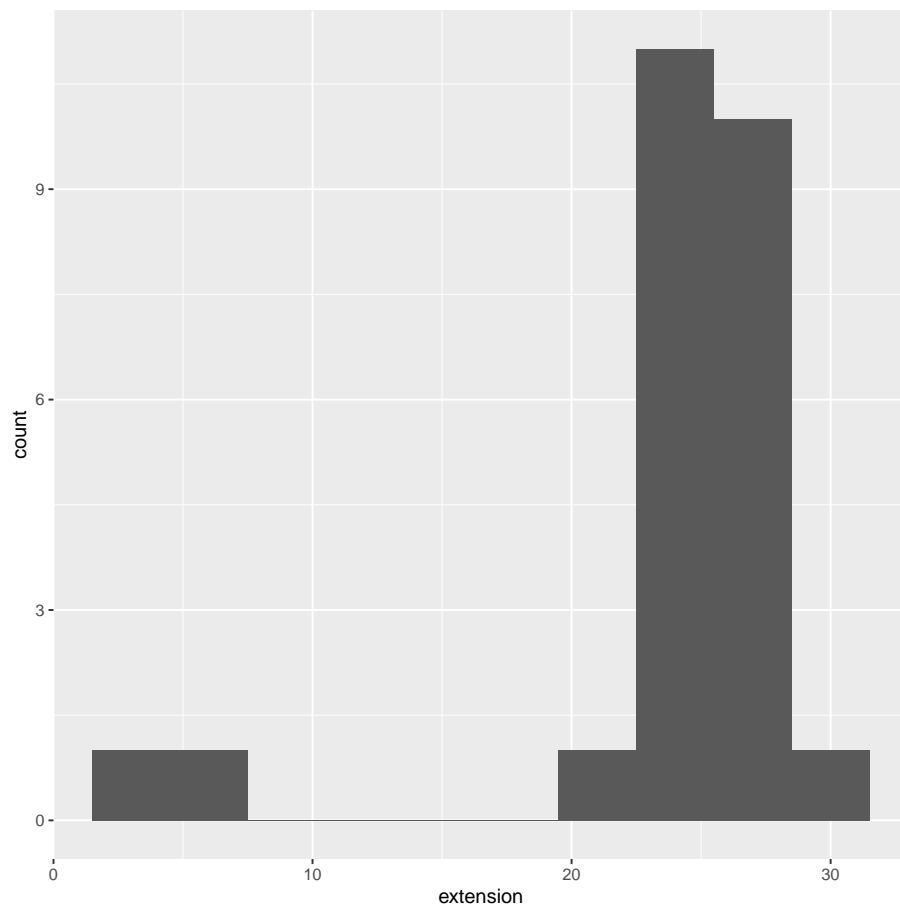


Figure 9: Histogram of wrist-extension data

```

t.test(wrist$extension, mu=24, alternative="greater")

##
## One Sample t-test
##
## data:  wrist$extension
## t = 0.2011, df = 24, p-value = 0.4212
## alternative hypothesis: true mean is greater than 24
## 95 percent confidence interval:
##  22.19819      Inf
## sample estimates:
## mean of x
##    24.24

```

Figure 10: T-test for wrist-extension data

```

sign_test(wrist, extension, 24)

## $above_below
## below above
##    3    17
##
## $p_values
## alternative    p_value
## 1      lower 0.999798775
## 2      upper 0.001288414
## 3 two-sided 0.002576828

```

Figure 11: Sign test for wrist-extension data

Suppose we have a data frame like this:

```
d
## # A tibble: 5 x 2
##       x g
##   <dbl> <chr>
## 1    10 a
## 2    11 b
## 3    12 a
## 4    13 b
## 5    14 a
```

Then we can select all the rows for which the column `g` is equal to the text `a` like this:

```
d %>% filter(g=="a")
## # A tibble: 3 x 2
##       x g
##   <dbl> <chr>
## 1    10 a
## 2    12 a
## 3    14 a
```

Figure 12: Example of “filter”

```

wrist %>% filter(extension>10) -> wrist2
t.test(wrist2$extension, mu=24, alternative="greater")

##
## One Sample t-test
##
## data: wrist2$extension
## t = 4.5189, df = 22, p-value = 8.492e-05
## alternative hypothesis: true mean is greater than 24
## 95 percent confidence interval:
## 25.15915      Inf
## sample estimates:
## mean of x
## 25.86957

sign_test(wrist2, extension, 24)

## $above_below
## below above
##      1      17
##
## $p_values
##   alternative      p_value
## 1      lower 9.999962e-01
## 2      upper 7.247925e-05
## 3 two-sided 1.449585e-04

```

Figure 13: Further analysis for wrist extension data

```

## # A tibble: 12 x 4
##   student pre_test post_test difference
##   <dbl>   <dbl>   <dbl>   <dbl>
## 1     1     510     850     340
## 2     2     610     790     180
## 3     3     640     850     210
## 4     4     675     775     100
## 5     5     600     700     100
## 6     6     550     775     225
## 7     7     610     700     90
## 8     8     625     850     225
## 9     9     450     690     240
## 10    10     720     775     55
## 11    11     575     540    -35
## 12    12     675     680     5

```

Figure 14: Chess study data

```

hsam
## # A tibble: 29 x 2
##   memory test_score
##   <chr>     <dbl>
## 1 control     4
## 2 control     4
## 3 control     3
## 4 hsam        6
## 5 control     4
## 6 control     6
## 7 control     5
## 8 hsam        7
## 9 hsam        6
## 10 hsam       4
## # ... with 19 more rows

```

Figure 15: HSAM data

```
hsam %>% summarize(m=median(test_score))  
  
## # A tibble: 1 x 1  
##       m  
##   <dbl>  
## 1     4
```

Figure 16: HSAM overall median test score

```
## $table  
##       above  
## group  above below  
## control    5    10  
## hsam       8     0  
##  
## $test  
##       what      value  
## 1 statistic 9.435897436  
## 2         df 1.000000000  
## 3   P-value 0.002127789
```

Figure 17: HSAM Mood's median test

```
ggplot(hsam, aes(sample=test_score)) +  
  stat_qq() + stat_qq_line() + facet_wrap(~memory)
```

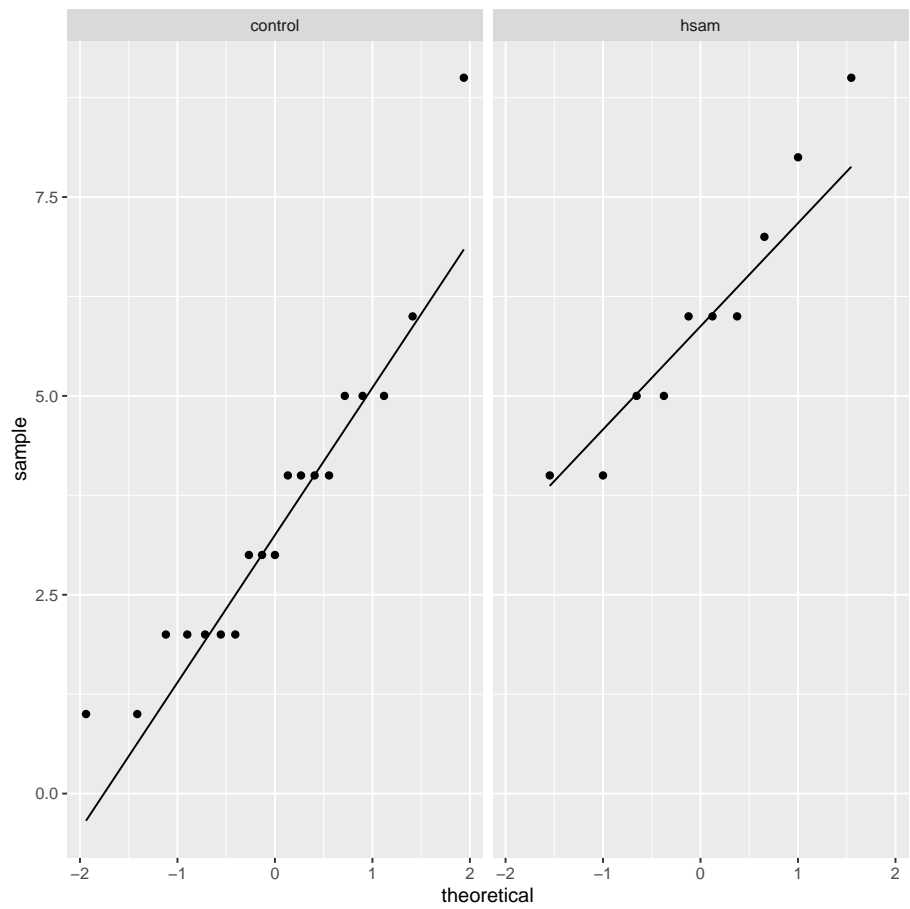


Figure 18: HSAM faceted normal quantile plots

```

## Parsed with column specification:
## cols(
##   plot = col_character(),
##   penetrability = col_double()
## )

## # A tibble: 20 x 2
##   plot penetrability
##   <chr>      <dbl>
## 1 B          3.13
## 2 C          4.91
## 3 B          3.26
## 4 A          3
## 5 B          3.86
## 6 A          2.9
## 7 A          2.86
## 8 C          3.99
## 9 B          3.38
## 10 A         3.18
## 11 C         4.3
## 12 C         3.94
## 13 A         2.92
## 14 A         2.86
## 15 B         3.38
## 16 C         4.2
## 17 A         2.96
## 18 B         3.02
## 19 C         4.34
## 20 A         2.78

```

Figure 19: Penetrability data (randomly chosen rows)

Figure 20: Penetrability data boxplot

```

ggplot(soil, aes(sample=penetrability)) + stat_qq() +
  stat_qq_line() + facet_wrap(~plot, ncol=2)

```

The purpose of the `ncol=2` is to arrange the plots as three cells of a  $2 \times 2$  grid. By default, the three plots will come out in one row, side by side, which is harder to read.

Figure 21: Penetrability data faceted normal quantile plots



Part (i):

```
soil.1=aov(penetrability~plot, data=soil)
summary(soil.1)

##              Df Sum Sq Mean Sq F value Pr(>F)
## plot          2 18.260    9.130  140.5 <2e-16 ***
## Residuals    57  3.703    0.065
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Part (ii):

```
TukeyHSD(soil.1)

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = penetrability ~ plot, data = soil)
##
## $plot
##      diff      lwr      upr    p adj
## B-A 0.428 0.2340375 0.6219625 5.6e-06
## C-A 1.324 1.1300375 1.5179625 0.0e+00
## C-B 0.896 0.7020375 1.0899625 0.0e+00
```

Figure 22: Penetrability data ANOVA

Part (i):

```
oneway.test(penetrability~plot, data=soil)

##
## One-way analysis of means (not assuming equal variances)
##
## data: penetrability and plot
## F = 186.86, num df = 2.000, denom df = 33.243, p-value < 2.2e-16
```

Part (ii):

```
gamesHowellTest(penetrability~factor(plot), data=soil)

##
## Pairwise comparisons using Games-Howell test
## data: penetrability by factor(plot)

## A B
## B 2.7e-05 -
## C < 2e-16 4.5e-11

##
## P value adjustment method: none
## alternative hypothesis: two.sided
```

Figure 23: Penetrability data Welch ANOVA

Part (i):

```
median_test(soil, penetrability, plot)

## $table
##      above
## group above below
##   A      0      20
##   B     10     10
##   C     20      0
##
## $test
##      what      value
## 1 statistic 4.000000e+01
## 2          df 2.000000e+00
## 3   P-value 2.061154e-09
```

Part (ii):

```
pairwise_median_test(soil, penetrability, plot)

## # A tibble: 3 x 4
##   g1    g2    p_value adj_p_value
##   <chr> <chr>    <dbl>    <dbl>
## 1 A     B     5.40e- 6    1.62e- 5
## 2 A     C     2.54e-10    7.62e-10
## 3 B     C     1.25e- 8    3.76e- 8
```

Figure 24: Penetrability data Mood's median test