

Mixed-Factorial ANOVA Demo

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2025-04-01

Mixed Factorial ANOVA | Research Context Prompt

The same stress researcher from our previous demonstration conducted another study to determine the impact of psychosocial stress on participants processing speed ability using a mixed-factorial research design.

The researcher randomly assigned 30 participants to one of two stress conditions: (1) Low Stress and (2) High Stress. Participants were assessed at three separate times across their academic career (Sophomore, Junior, and Senior) measuring a new processing speed performance measure. The dependent measure in this task was processing speed ability measured in milliseconds (500 ms = .5 seconds).

Again, this researcher believed that regulating feelings of stress would impair concurrent performance of processing speed and that these effects would materialize across time.

Mixed Factorial ANOVA | Data

```
library(psych)
library(tidyverse)
library(jmv)
library(ggpubr)
library(apaTables)
library(ez)
library(rstatix)
library(reshape)

dat_MF <- read.csv("Mixed_Factorial.csv")
dat_MF$Time <- 1000 - dat_MF$Time

# SUPER IMPORTANT FOR WITHIN SUBJECT DESIGNS
# Set Subject or ID variable to a factor
dat_MF$Subject <- factor(dat_MF$Subject)

# Set IVs to factors
dat_MF$Condition <- factor(dat_MF$Condition, labels = c("Low-Stress", "High-Stress"))
dat_MF$Rank <- factor(dat_MF$Rank, labels = c("Sophomore", "Junior", "Senior"))

# Assess number of participants and variables in data set
dim(dat_MF)
```

```
## [1] 90  4
```

```
head(dat_MF)
```

##	Subject	Condition	Rank	Time
## 1	1	Low-Stress	Sophomore	488
## 2	1	Low-Stress	Junior	459
## 3	1	Low-Stress	Senior	397
## 4	2	High-Stress	Sophomore	490
## 5	2	High-Stress	Junior	446
## 6	2	High-Stress	Senior	404

```
# To use anovaRM() we must reshape our data set from `long` to `wide`
dat_MFW <- cast(dat_MF,
                Subject + Condition ~ Rank, #Isolates each ID and Condition to form columns for
Rank
                value = "Time")           #Tells function to use Time variable to fill in Rank
s

dat_MFW$averageTime <- (dat_MFW$Sophomore + dat_MFW$Junior + dat_MFW$Senior) / 3 # This is used
later for assumption checking
```

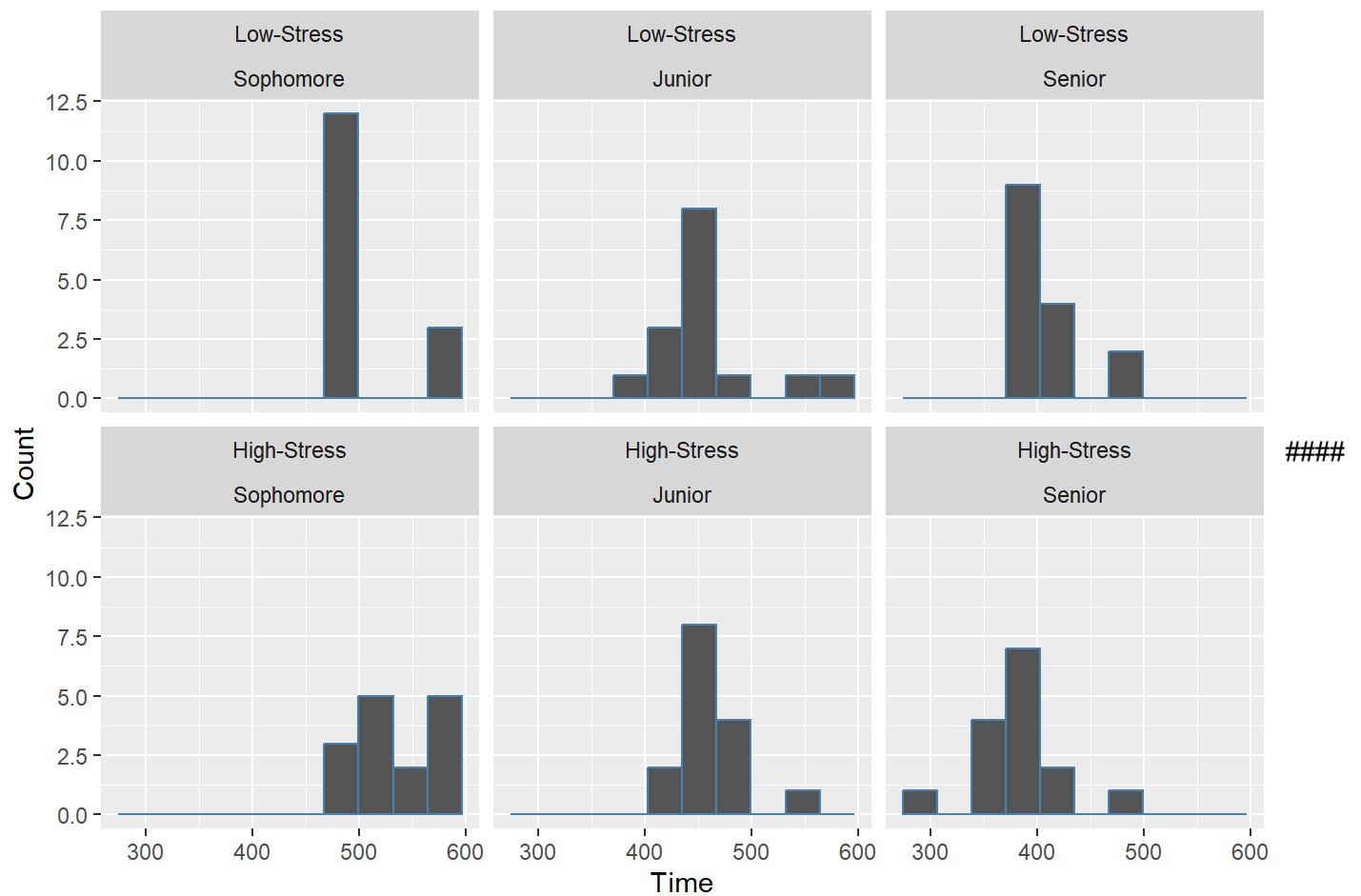
Mixed Factorial ANOVA | Descriptive Stats

```
# Descriptive Stats
describeBy(dat_MF$Time,
           list(dat_MF$Condition, dat_MF$Rank),
           mat=TRUE)
```

##	item	group1	group2	vars	n	mean	sd	median	trimmed
## X11	1	Low-Stress	Sophomore	1	15	503.2667	40.83672	488	499.4615
## X12	2	High-Stress	Sophomore	1	15	534.4000	36.12834	520	533.6923
## X13	3	Low-Stress	Junior	1	15	456.4000	46.65650	441	452.1538
## X14	4	High-Stress	Junior	1	15	464.1333	26.90955	463	460.8462
## X15	5	Low-Stress	Senior	1	15	412.6667	33.77164	402	408.5385
## X16	6	High-Stress	Senior	1	15	381.5333	42.05416	384	379.4615
##	mad	min	max	range	skew	kurtosis	se		
## X11	8.8956	468	588	120	1.2901196	-0.135849	10.543997		
## X12	44.4780	488	590	102	0.1840255	-1.659049	9.328298		
## X13	28.1694	400	568	168	1.1918834	0.467006	12.046655		
## X14	23.7216	431	540	109	1.2570249	1.603398	6.948016		
## X15	8.8956	382	497	115	1.7747683	1.588227	8.719800		
## X16	26.6868	300	490	190	0.5465190	1.044874	10.858337		

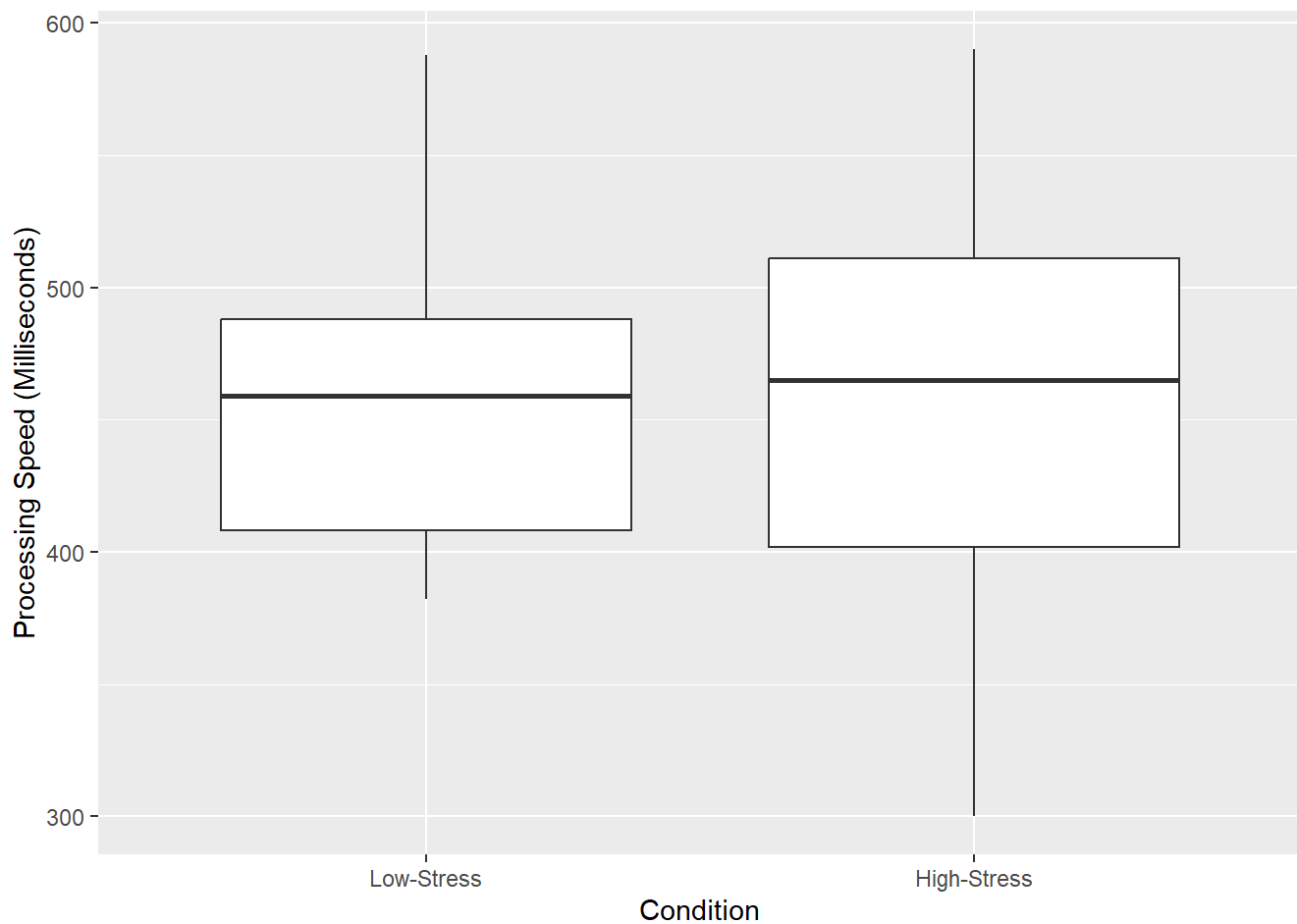
Mixed Factorial ANOVA | Histogram

```
ggplot(data = dat_MF,
       mapping = aes(x = Time)) +
  geom_histogram(bins = 10, color = "steelblue") +
  labs(y = "Count", x = "Time")+
  facet_wrap(~dat_MF$Condition + dat_MF$Rank)
```



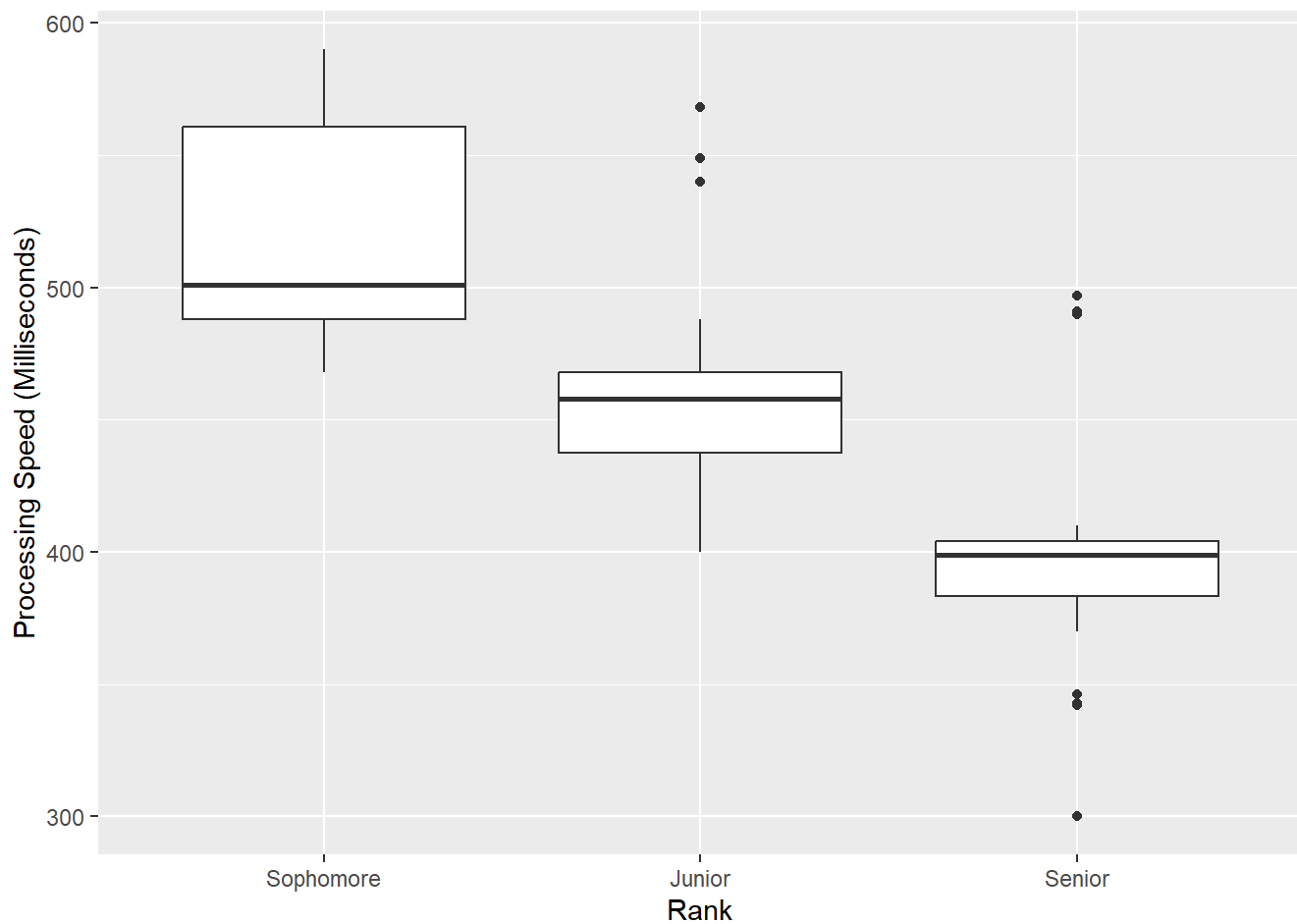
Mixed Factorial ANOVA | Boxplot

```
ggplot(data = dat_MF,
       mapping = aes(y = Time, x = Condition)) +
  geom_boxplot() +
  labs(y = "Processing Speed (Milliseconds)")
```



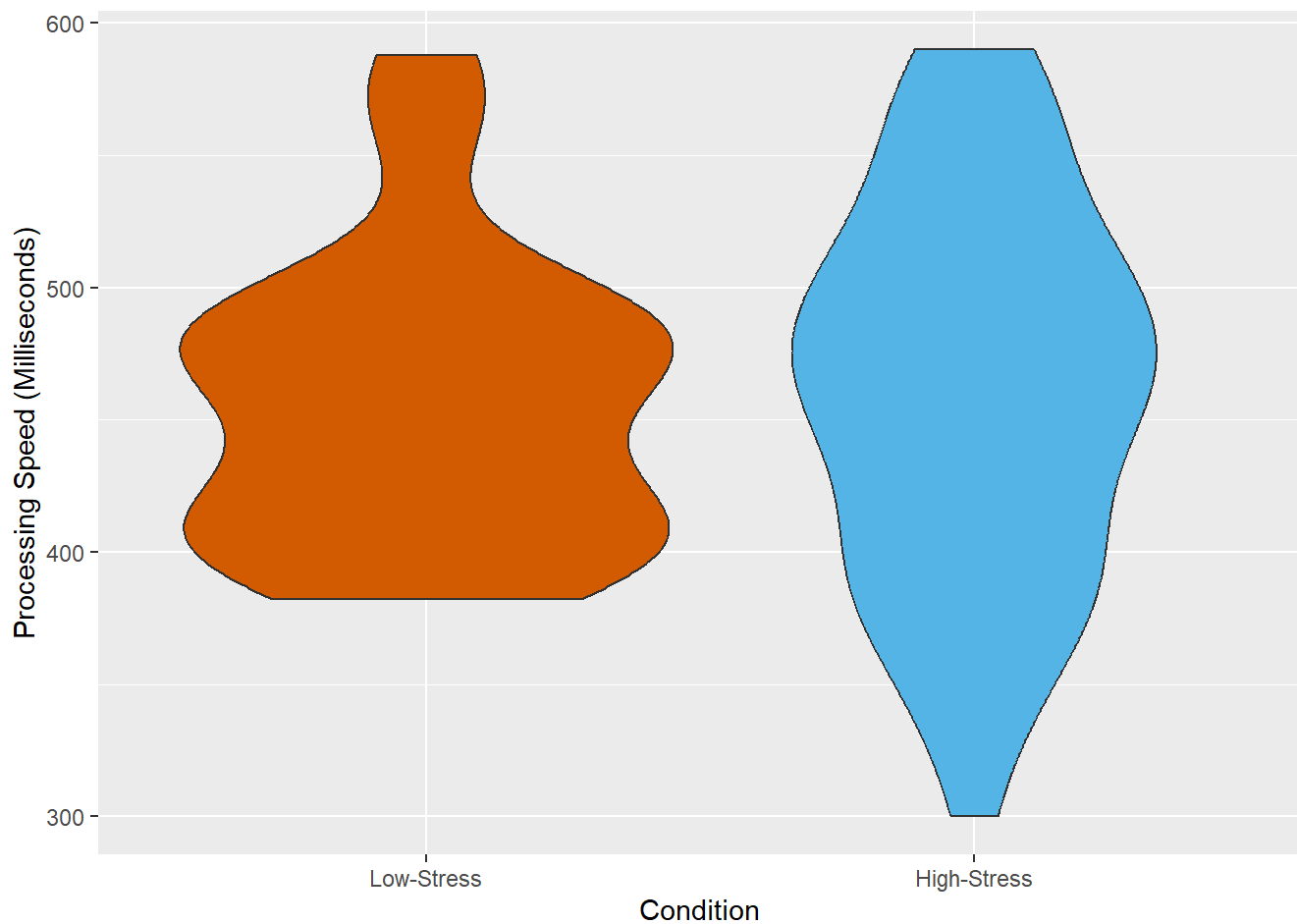
Mixed Factorial ANOVA | Boxplot

```
ggplot(data = dat_MF,  
       mapping = aes(y = Time, x = Rank)) +  
  geom_boxplot() +  
  labs(y = "Processing Speed (Milliseconds)")
```



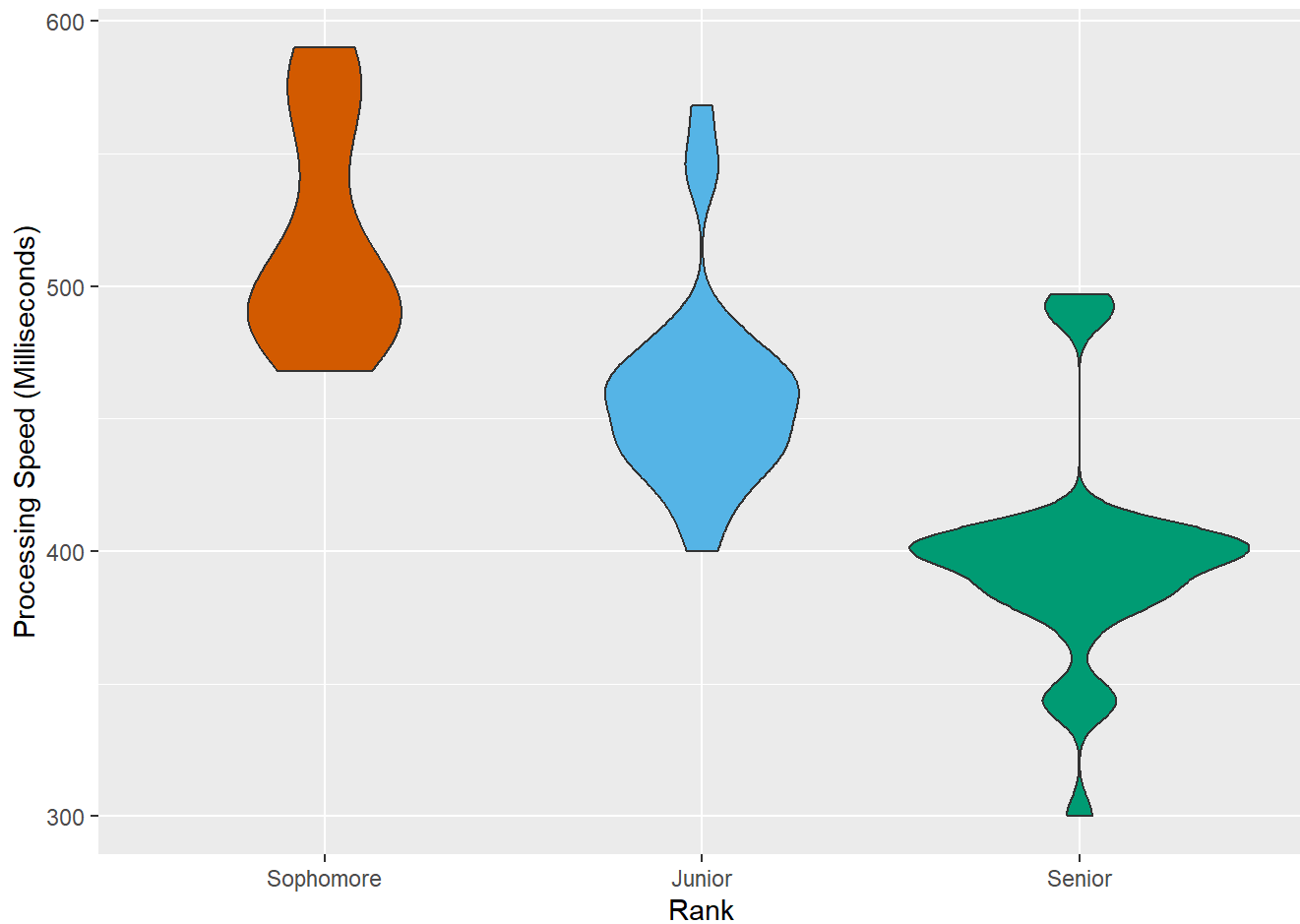
Mixed Factorial ANOVA | Violin Plot

```
ggplot(data = dat_MF,
       aes(x = Condition, y = Time,
           fill = Condition))+
  geom_violin(show.legend=FALSE)+
  ylab("Processing Speed (Milliseconds)")+
  xlab("Condition")+
  scale_fill_manual(values=c("#D55e00", "#56B4E9", "#009E73"))
```



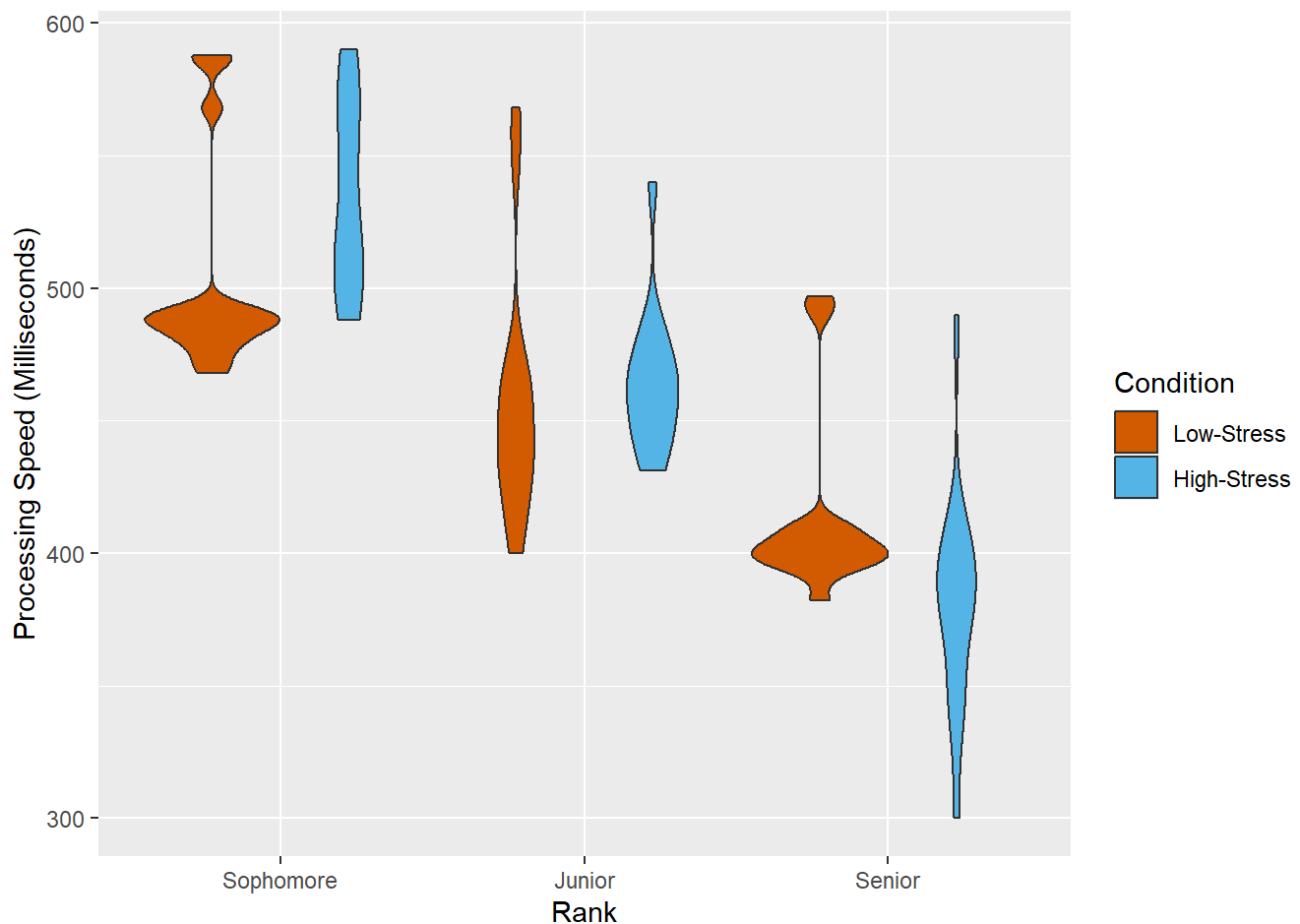
Mixed Factorial ANOVA | Violin Plot

```
ggplot(data = dat_MF,  
       aes(x = Rank, y = Time,  
           fill = Rank))+  
  geom_violin(show.legend=FALSE)+  
  ylab("Processing Speed (Milliseconds)") +  
  xlab("Rank") +  
  scale_fill_manual(values=c("#D55e00", "#56B4E9", "#009E73"))
```



Mixed Factorial ANOVA | Grouped Violin Plot

```
ggplot(data = dat_MF,  
       aes(x = Rank, y = Time,  
           fill = Condition))+  
geom_violin(show.legend=TRUE)+  
ylab("Processing Speed (Milliseconds)")+  
xlab("Rank")+  
scale_fill_manual(values=c("#D55e00", "#56B4E9", "#009E73"))
```



Mixed Factorial ANOVA | anovaRM() Omnibus Assumption Checks

```
anovaRM(data=dat_MFW,
  bs = list('Condition'),
  bsTerms = list('Condition'),
  rm = list(
    list(label = 'Rank',
      levels = list('Sophomore',
        'Junior',
        'Senior'))),
  rmCells = list(
    list(measure = 'Sophomore',
      cell = 'Sophomore'),
    list(measure = 'Junior',
      cell = 'Junior'),
    list(measure = 'Senior',
      cell = 'Senior')),
  rmTerms= list('Rank'),
  leveneTest = TRUE,
  qq = TRUE,
  spherTests = TRUE)
```



```
##
## REPEATED MEASURES ANOVA
##
## Within Subjects Effects
##
```

	Sum of Squares	df	Mean Square	F	p
Rank	222390.87	2	111195.4333	158.38122	< .0000001
Rank:Condition	14838.29	2	7419.1444	10.56746	0.0001277
Residual	39316.18	56	702.0746		

```
##
## Note. Type 3 Sums of Squares
##
##
## Between Subjects Effects
##
```

	Sum of Squares	df	Mean Square	F	p
Condition	149.5111	1	149.5111	0.05004870	0.8246028
Residual	83644.7556	28	2987.3127		

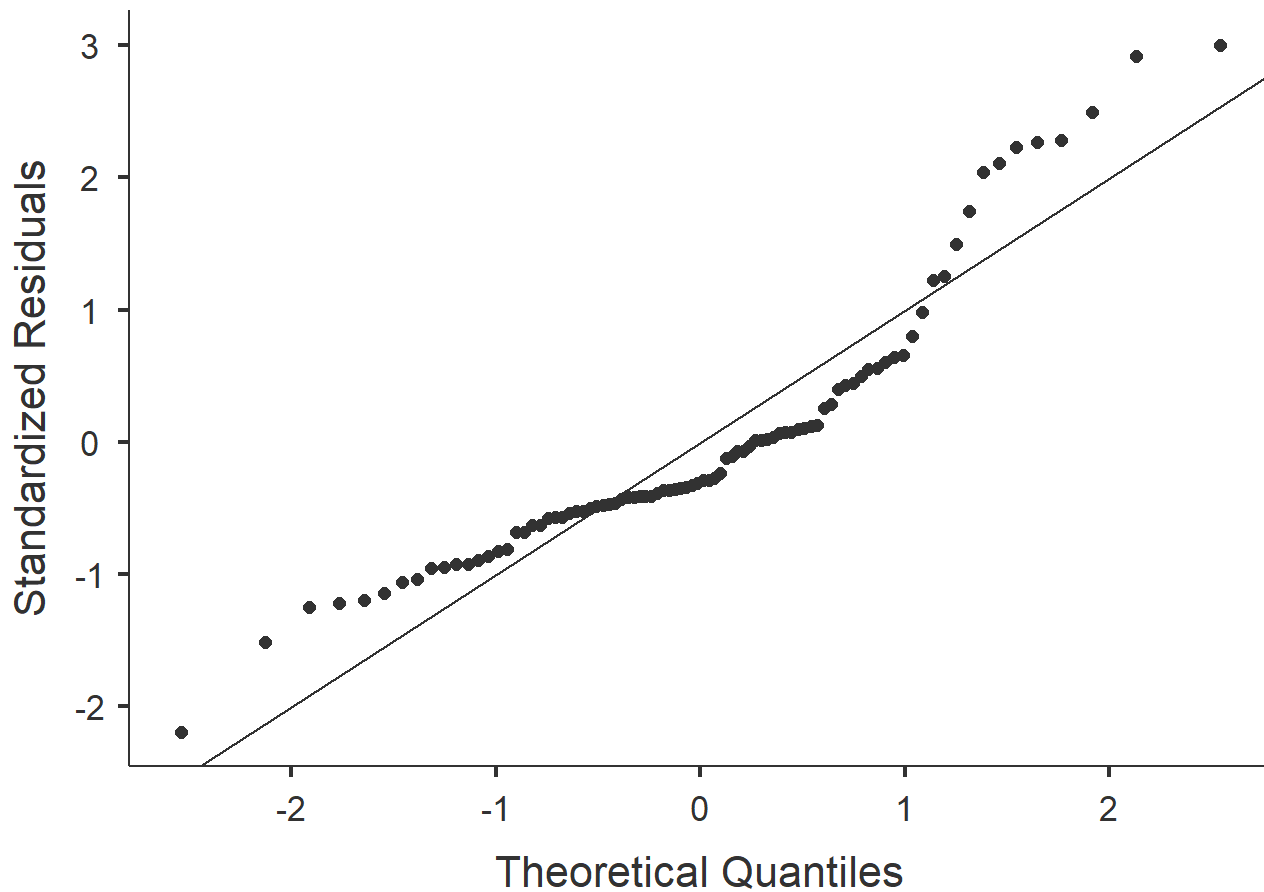
```
##
## Note. Type 3 Sums of Squares
##
##
## ASSUMPTIONS
##
## Tests of Sphericity
##
```

	Mauchly's W	p	Greenhouse-Geisser ϵ	Huynh-Feldt ϵ
Rank	0.6845379	0.0059961	0.7601891	0.7947804

```
##
##
## Homogeneity of Variances Test (Levene's)
##
```

	F	df1	df2	p
Sophomore	0.02578285	1	28	0.8735846
Junior	2.03565460	1	28	0.1647047
Senior	0.32438535	1	28	0.5735255

```
##
```



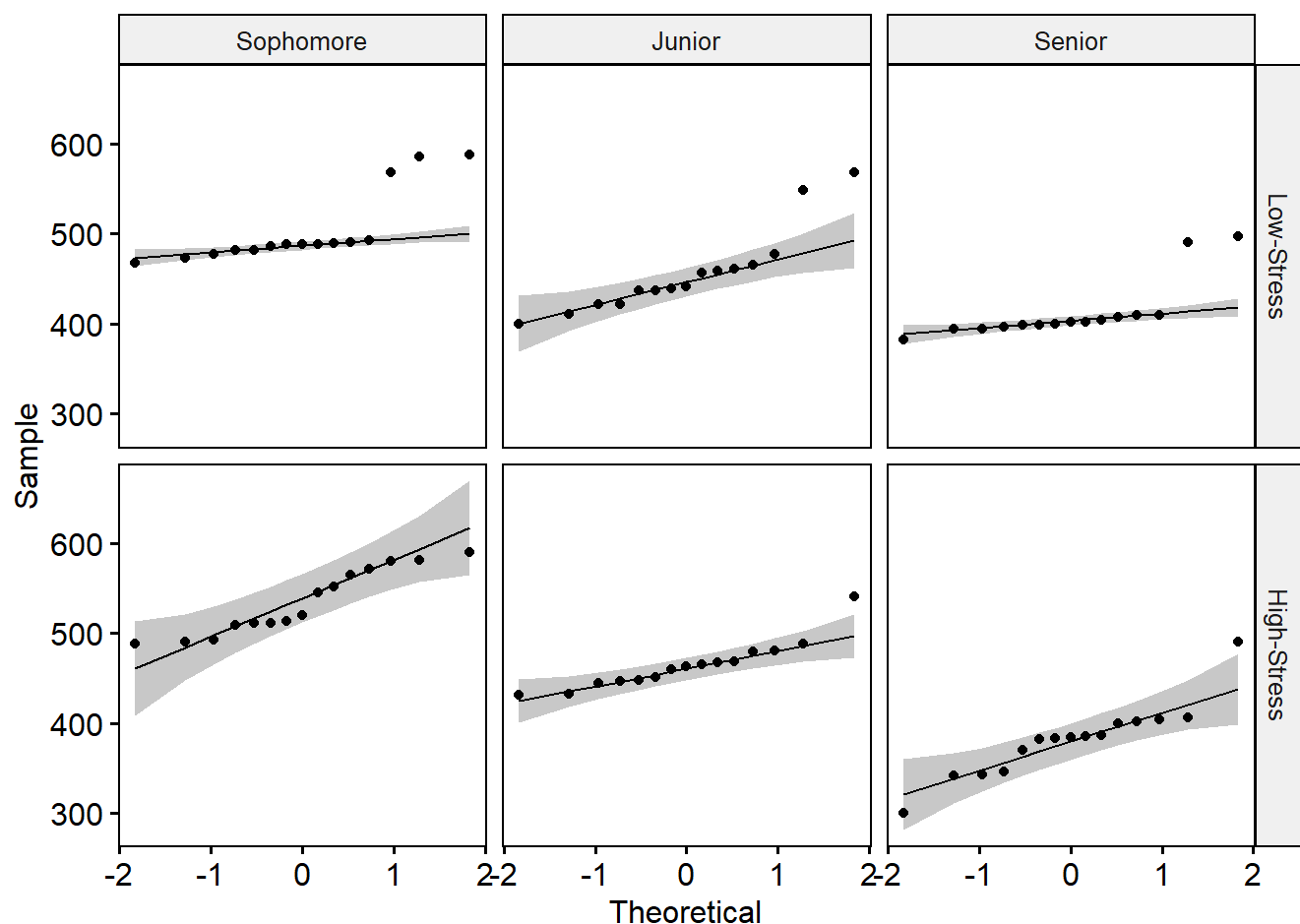
Mixed Factorial ANOVA | Group Level Assumption Checks

```
dat_MF %>%
  group_by(Condition, Rank) %>%
  shapiro_test(Time)
```

```
## # A tibble: 6 × 5
##   Condition Rank    variable statistic      p
##   <fct>      <fct>   <chr>      <dbl>    <dbl>
## 1 Low-Stress Sophomore Time      0.672 0.000126
## 2 Low-Stress Junior    Time      0.841 0.0131
## 3 Low-Stress Senior    Time      0.617 0.0000365
## 4 High-Stress Sophomore Time      0.901 0.0977
## 5 High-Stress Junior    Time      0.875 0.0395
## 6 High-Stress Senior    Time      0.899 0.0935
```

Mixed Factorial ANOVA | Group Level Assumption Checks

```
ggqqplot(dat_MF,
  'Time',
  facet.by = c("Condition", "Rank"))
```



Mixed Factorial ANOVA | Group Level Assumption Checks

```
dat_MFW %>%
  levene_test(averageTime ~ Condition,
              center = mean)
```

```
## # A tibble: 1 × 4
##   df1  df2 statistic    p
##   <int> <int>   <dbl> <dbl>
## 1     1    28     1.90 0.179
```

Mixed Factorial ANOVA | Group Level Assumption Checks

Must be conducted ON A MATRIX!! Also, note that it is using the Long version of the data!

```
box_m(as.data.frame(dat_MF$Time),
      dat_MF$Condition)
```

```
## # A tibble: 1 × 4
##   statistic p.value parameter method
##   <dbl>   <dbl>   <dbl> <chr>
## 1     3.31 0.0690         1 Box's M-test for Homogeneity of Covariance Matric...
```

Mixed Factorial ANOVA | Conducting the ANOVA

```
options(digits = 3)
anovaRM(data = dat_MFW,
        bs = list('Condition'),
        bsTerms = list('Condition'),
        rm = list(
            list(label = 'Rank',
                 levels = list('Sophomore',
                               'Junior',
                               'Senior'))),
        rmCells = list(
            list(measure = 'Sophomore',
                 cell = 'Sophomore'),
            list(measure = 'Junior',
                 cell = 'Junior'),
            list(measure = 'Senior',
                 cell = 'Senior')),
        rmTerms = list('Rank'),
        postHoc = list('Rank', c('Rank', 'Condition')),
        postHocCorr = 'bonf',
        emMeans = list('Rank', 'Condition', c('Rank', 'Condition')),
        emmPlots = TRUE,
        emmTables = TRUE,
        effectSize = 'partEta')
```

```
##
## REPEATED MEASURES ANOVA
##
## Within Subjects Effects
##
```

	Sum of Squares	df	Mean Square	F	p	η^2 -p
Rank	222391	2	111195	158.4	< .001	0.850
Rank:Condition	14838	2	7419	10.6	< .001	0.274
Residual	39316	56	702			

```
##
## Note. Type 3 Sums of Squares
##
##
## Between Subjects Effects
##
```

	Sum of Squares	df	Mean Square	F	p	η^2 -p
Condition	150	1	150	0.0500	0.825	0.002
Residual	83645	28	2987			

```
##
## Note. Type 3 Sums of Squares
##
##
## POST HOC TESTS
##
## Post Hoc Comparisons - Rank
##
```

Rank	Rank	Mean Difference	SE	df	t	p-bonferroni
Sophomore	- Junior	58.6	5.62	28.0	10.4	< .001
	- Senior	121.7	8.54	28.0	14.2	< .001
Junior	- Senior	63.2	5.98	28.0	10.6	< .001

```
##
##
## Post Hoc Comparisons - Rank:Condition
##
```

Rank	Condition	Rank	Condition	Mean Difference	SE	d
f	t	p-bonferroni				
8.0	Sophomore	Low-Stress	- Sophomore	High-Stress	-31.13	14.08
	-2.211	0.530	- Junior	Low-Stress	46.87	7.95
8.0	5.893	< .001	- Junior	High-Stress	39.13	13.99
8.0	2.797	0.138	- Senior	Low-Stress	90.60	12.08
8.0	7.499	< .001	- Senior	High-Stress	121.73	14.00
						2

8.0	8.694	< .001						
##		High-Stress	-	Junior	Low-Stress	78.00	13.99	2
8.0	5.574	< .001						
##			-	Junior	High-Stress	70.27	7.95	2
8.0	8.835	< .001						
##			-	Senior	Low-Stress	121.73	14.00	2
8.0	8.694	< .001						
##			-	Senior	High-Stress	152.87	12.08	2
8.0	12.653	< .001						
##	Junior	Low-Stress	-	Junior	High-Stress	-7.73	13.91	2
8.0	-0.556	1.000						
##			-	Senior	Low-Stress	43.73	8.46	2
8.0	5.168	< .001						
##			-	Senior	High-Stress	74.87	13.92	2
8.0	5.380	< .001						
##		High-Stress	-	Senior	Low-Stress	51.47	13.92	2
8.0	3.698	0.014						
##			-	Senior	High-Stress	82.60	8.46	2
8.0	9.761	< .001						
##	Senior	Low-Stress	-	Senior	High-Stress	31.13	13.93	2
8.0	2.236	0.503						
##								

##

##

ESTIMATED MARGINAL MEANS

##

RANK

##

Estimated Marginal Means - Rank

##

##	Rank	Mean	SE	Lower	Upper
##	Sophomore	519	7.04	504	533
##	Junior	460	6.95	446	475
##	Senior	397	6.96	383	411

##

##

CONDITION

##

Estimated Marginal Means - Condition

##

##	Condition	Mean	SE	Lower	Upper
##	Low-Stress	457	8.15	441	474
##	High-Stress	460	8.15	443	477

##

##

RANK:CONDITION

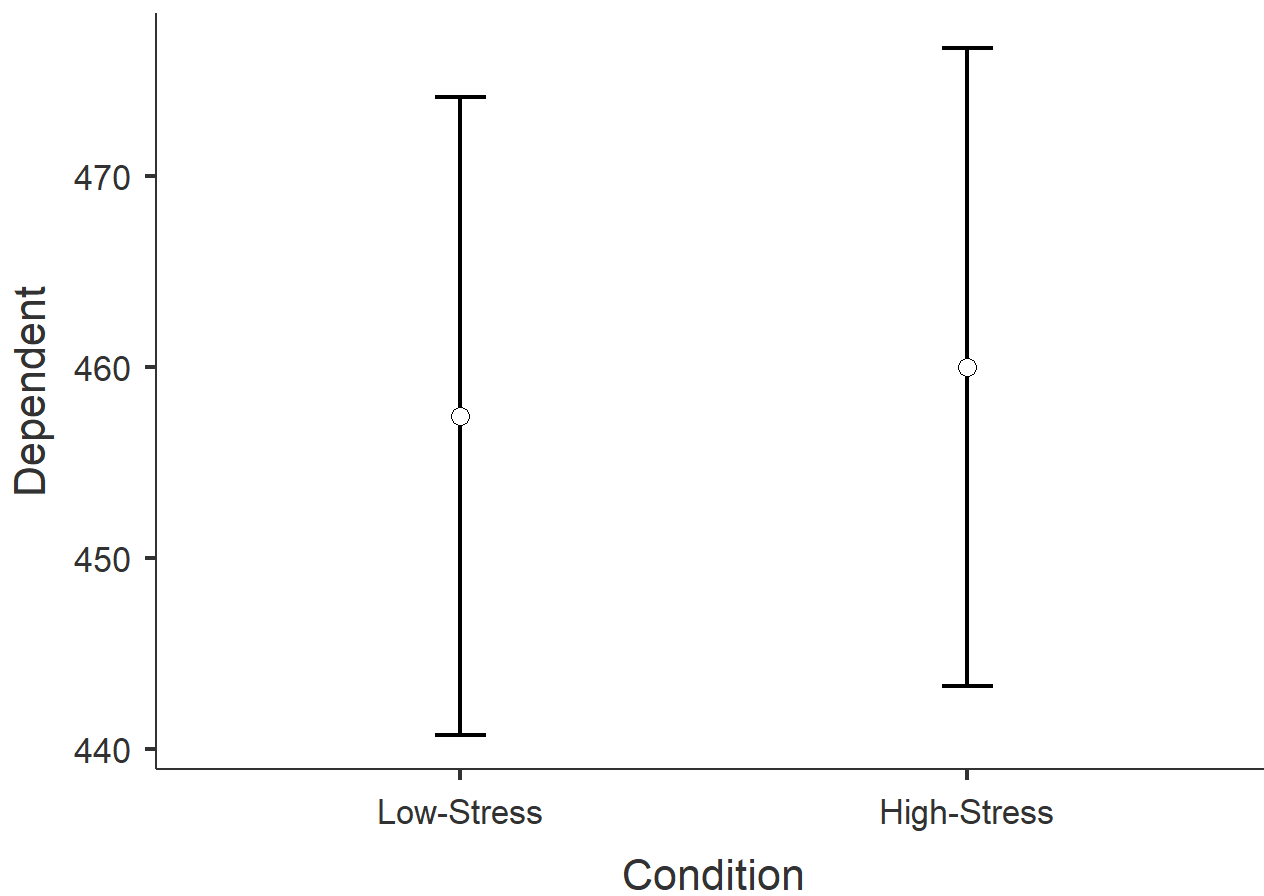
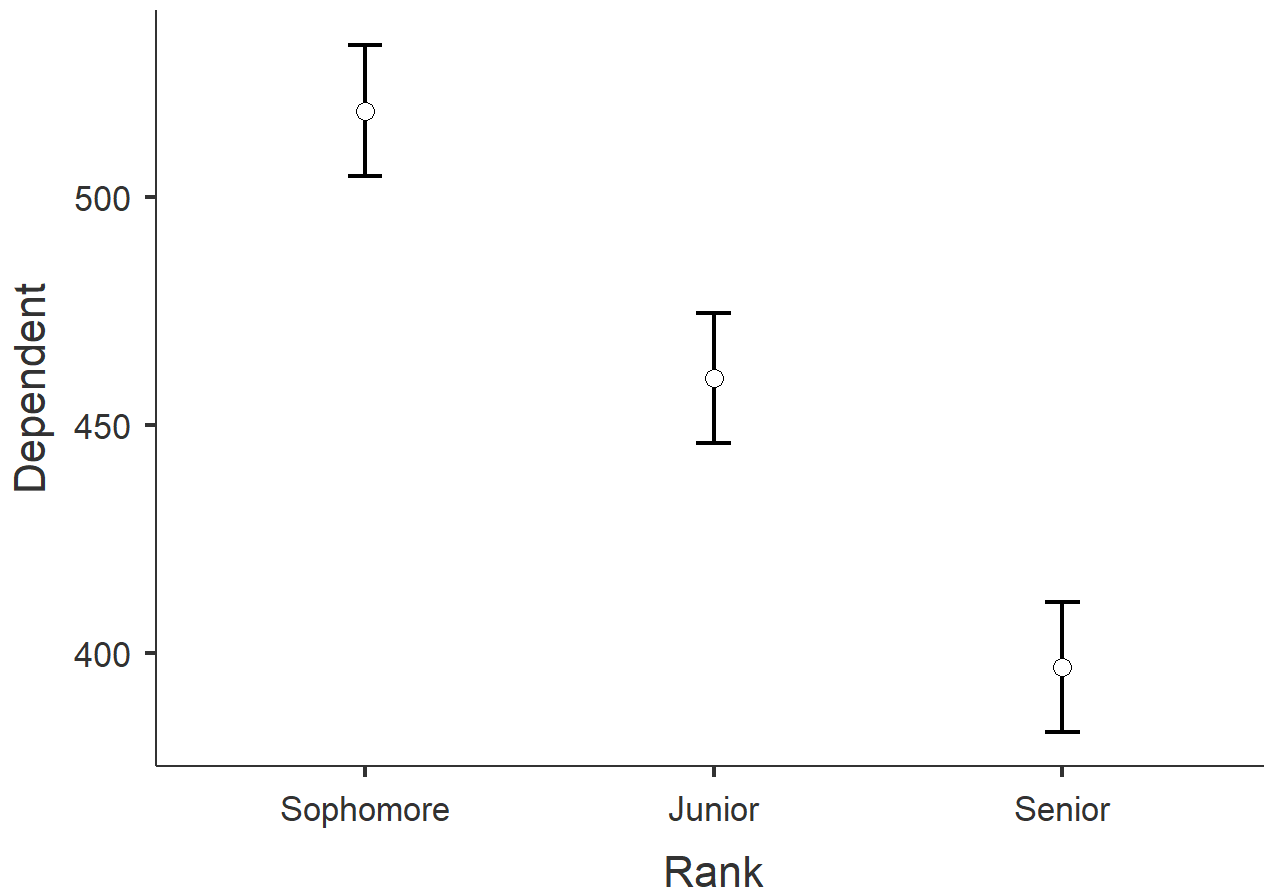
##

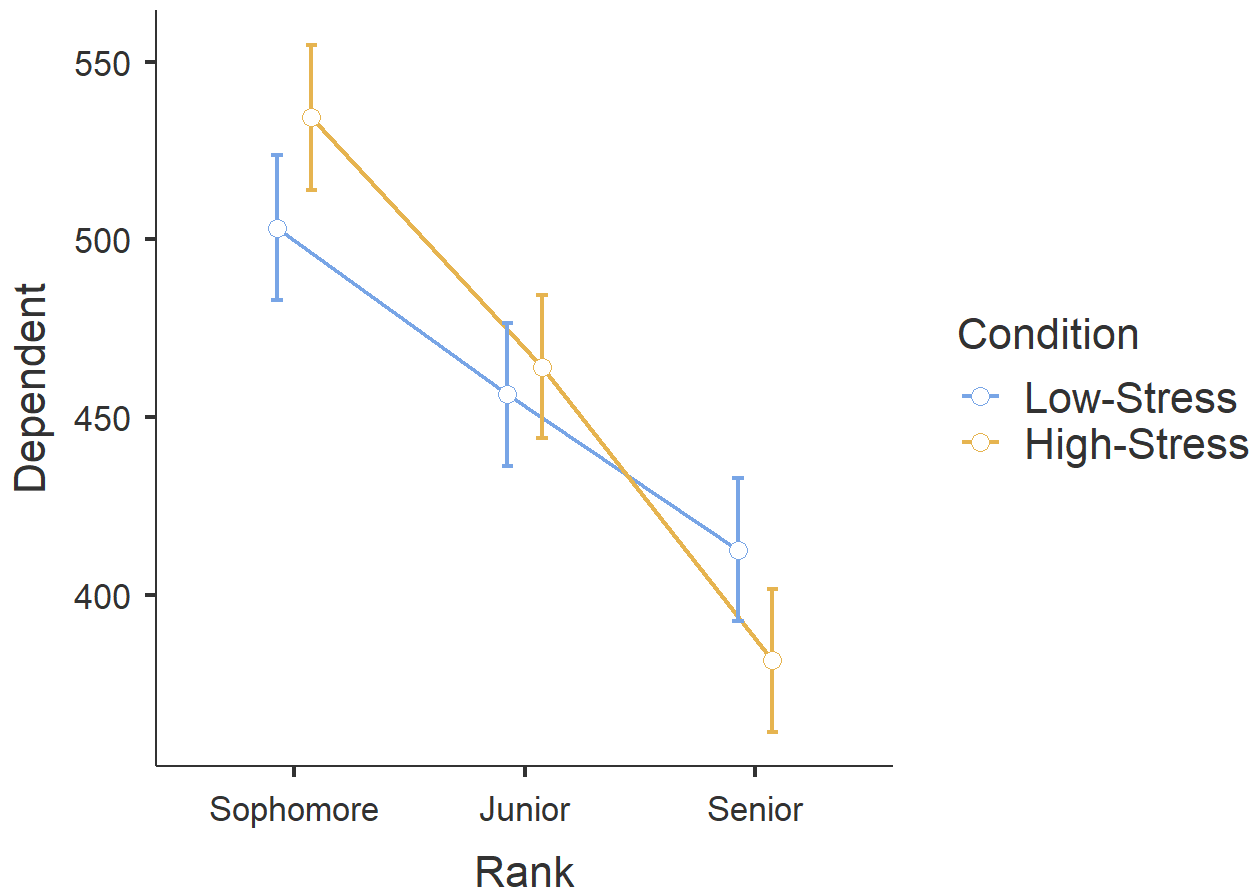
```
## Estimated Marginal Means - Rank:Condition
```

```
##
```

##	Condition	Rank	Mean	SE	Lower	Upper
##	<hr/>					
##	Low-Stress	Sophomore	503	9.95	483	524
##		Junior	456	9.83	436	477
##		Senior	413	9.85	392	433
##	High-Stress	Sophomore	534	9.95	514	555
##		Junior	464	9.83	444	484
##		Senior	382	9.85	361	402
##	<hr/>					

```
##
```





Mixed Factorial ANOVA | ANOVA Output: Interaction Simple Effects

```
one.way <- dat_MF %>%
  group_by(Condition) %>%
  anova_test(dv = Time,
             wid = Subject,
             within = Rank,
             effect.size = 'ges') %>%
  get_anova_table() %>%
  adjust_pvalue(method = "bonferroni")
```

one.way

```
## # A tibble: 2 x 9
##   Condition Effect   DFn  DFd    F      p `p<.05` ges   p.adj
##   <fct>      <chr> <dbl> <dbl> <dbl> <dbl> <chr> <dbl> <dbl>
## 1 Low-Stress Rank     2    28  86.5 1.04e-12 *   0.469 2.08e-12
## 2 High-Stress Rank    1.33 18.6  83.8 4.9 e- 9 *   0.768 9.8 e- 9
```

```
one.way2 <- dat_MF %>%
  group_by(Rank) %>%
  anova_test(dv = Time,
             between = Condition,
             effect.size = 'ges') %>%
  get_anova_table() %>%
  adjust_pvalue(method = "bonferroni")
```

```
one.way2
```

```
## # A tibble: 3 × 9
##   Rank      Effect    DFn  DFd    F      p `p<.05`    ges p.adj
##   <fct>    <chr>    <dbl> <dbl> <dbl> <dbl> <chr>    <dbl> <dbl>
## 1 Sophomore Condition     1    28 4.89  0.035 "*"      0.149 0.105
## 2 Junior    Condition     1    28 0.309 0.583 ""       0.011 1
## 3 Senior    Condition     1    28 5.00  0.034 "*"      0.151 0.102
```

Demonstrates that there are significant differences between the Ranks for participants in the Low-Stress [$F(2, 28) = 86.50, p < .001$] and High-Stress [$F(1.33, 18.64) = 83.79, p < .001$] conditions—furthermore, there is an issue of Sphericity demonstrated for the High-Stress condition (note the degrees of freedom!).

Mixed Factorial ANOVA | Professional ANOVA Visualization Code

Generates Marginal Means for Visualization--the p-values here are more appropriate to interpret compared to the chunk above!

```
pair <- dat_MF %>%
  group_by(Rank) %>%
  pairwise_t_test(Time~Condition,
                  p.adjust.method = "bonferroni") %>%
  add_xy_position(x = "Rank",
                  group = "Condition")
```

```
pair
```

```
## # A tibble: 3 × 15
##   Rank      .y. group1 group2    n1    n2      p p.signif p.adj p.adj.signif
##   <fct>    <chr> <chr>  <chr>  <int> <int>  <dbl> <chr>    <dbl> <chr>
## 1 Sophomore Time Low-St... High-...    15    15 0.0353 *      0.0353 *
## 2 Junior    Time Low-St... High-...    15    15 0.583 ns      0.583 ns
## 3 Senior    Time Low-St... High-...    15    15 0.0335 *      0.0335 *
## # i 5 more variables: y.position <dbl>, groups <named list>, x <dbl>,
## #   xmin <dbl>, xmax <dbl>
```

```
pair2 <- dat_MF %>%
  group_by(Condition) %>%
  pairwise_t_test(Time~Rank,
                  p.adjust.method = "bonferroni") %>%
  add_xy_position(x = "Condition",
                 group = "Rank")
```

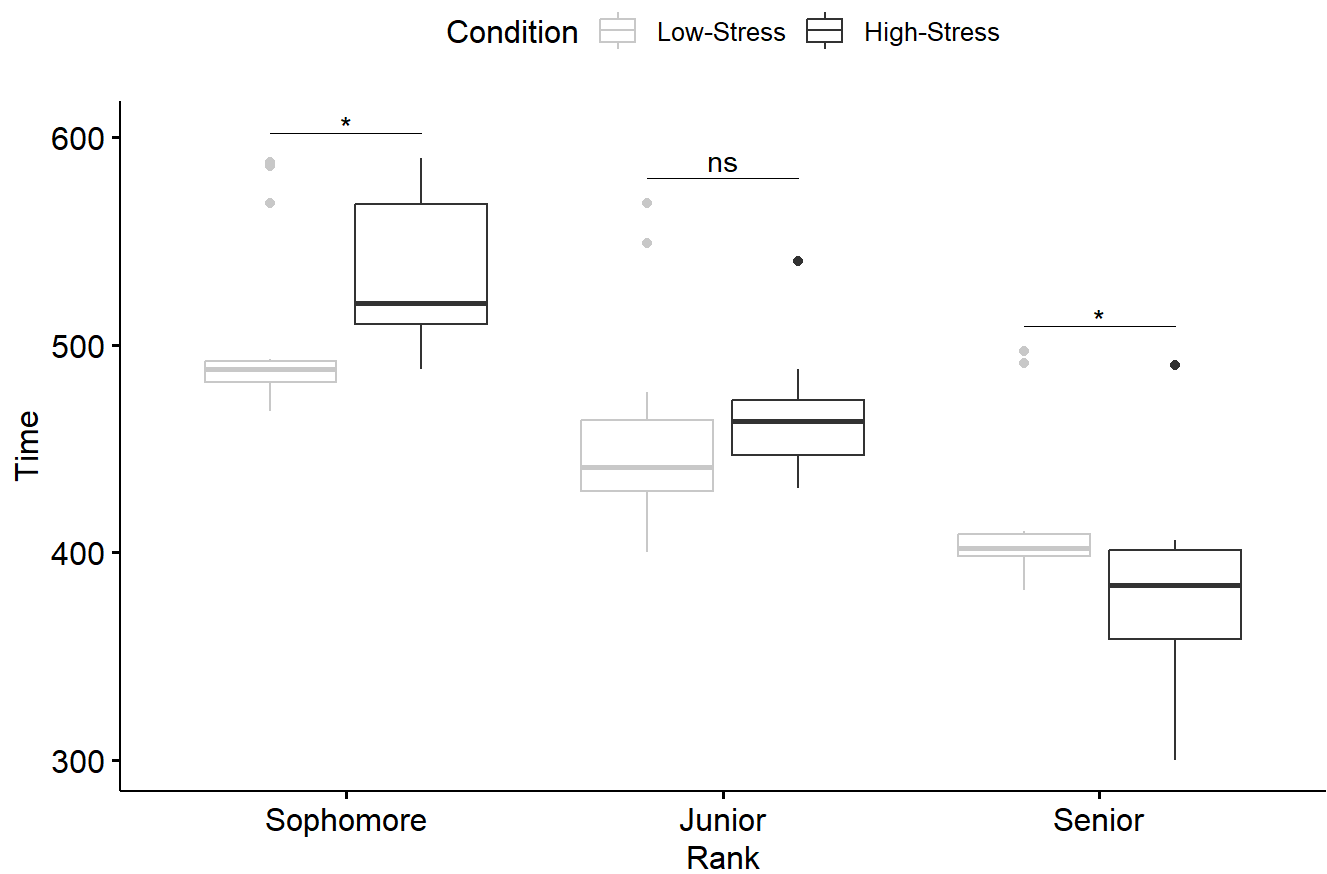
```
pair2
```

```
## # A tibble: 6 × 15
##   Condition .y. group1 group2 n1 n2 p p.signif p.adj
##   <fct>      <chr> <chr>   <chr> <int> <int> <dbl> <chr>    <dbl>
## 1 Low-Stress Time Sophomore Junior 15 15 3.02e- 3 ** 9.05e- 3
## 2 Low-Stress Time Sophomore Senior 15 15 2.98e- 7 **** 8.94e- 7
## 3 Low-Stress Time Junior Senior 15 15 5.34e- 3 ** 1.6 e- 2
## 4 High-Stress Time Sophomore Junior 15 15 2.8 e- 6 **** 8.41e- 6
## 5 High-Stress Time Sophomore Senior 15 15 7.04e-15 **** 2.11e-14
## 6 High-Stress Time Junior Senior 15 15 1.21e- 7 **** 3.64e- 7
## # i 6 more variables: p.adj.signif <chr>, y.position <dbl>,
## # groups <named list>, x <dbl>, xmin <dbl>, xmax <dbl>
```

Demonstrates that the difference between high- and low-stress conditions only materialized when participants were in their Sophomore ($p = .04$) and Senior ($p = .03$) years, but not when they were Juniors ($p = .58$)

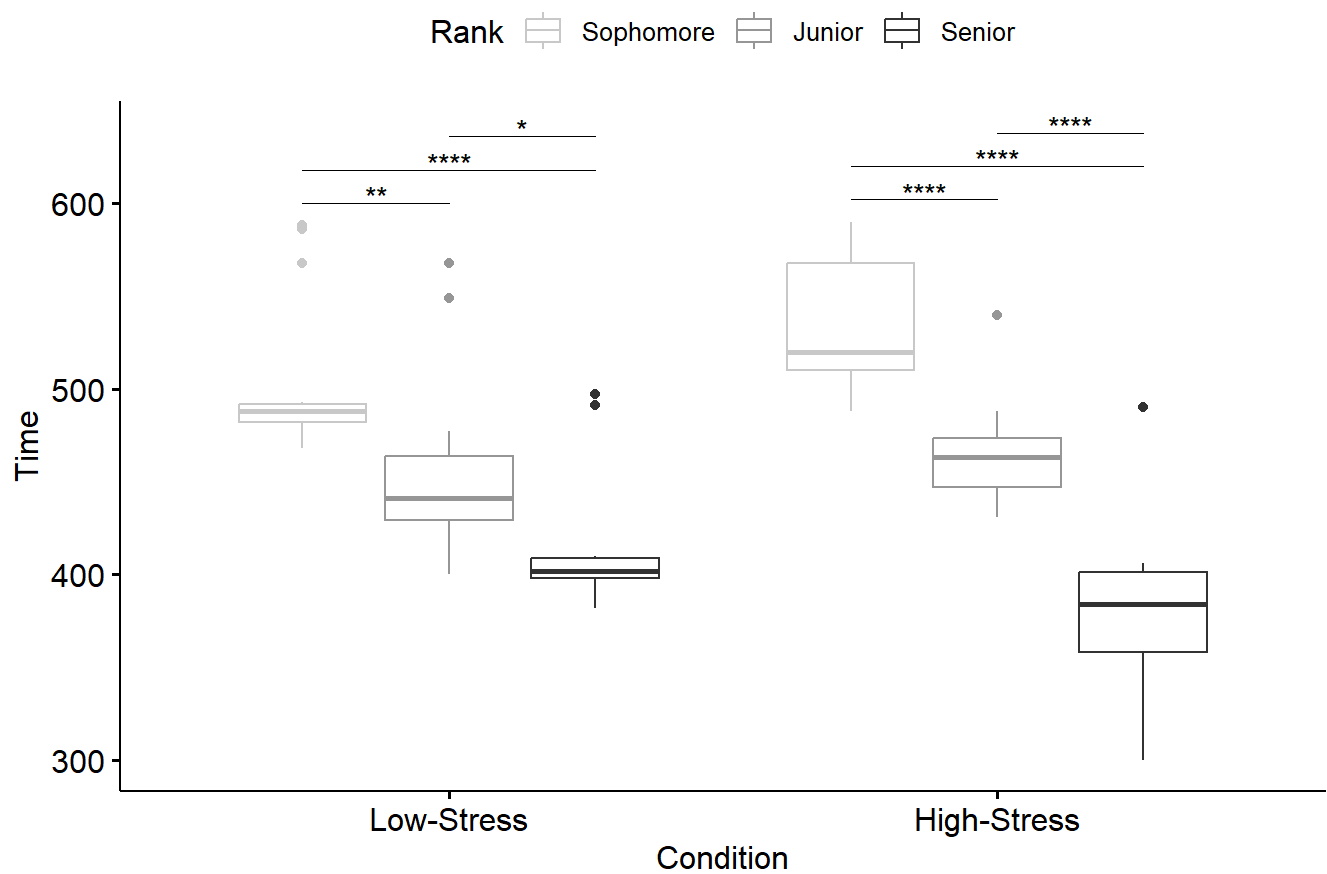
Mixed Factorial ANOVA | Professional ANOVA Visualization Code

```
ggboxplot(dat_MF,
          x = "Rank",
          y = "Time",
          color = "Condition",
          palette = "grey")+
  stat_pvalue_manual(pair,
                    hide.ns = FALSE,
                    tip.length = 0)+
  labs(
    caption = get_pwc_label(pair)
  )
```



pwc: **T test**; p.adjust: **Bonferroni**

```
ggboxplot(dat_MF,
  x = "Condition",
  y = "Time",
  color = "Rank",
  palette = "grey")+
stat_pvalue_manual(pair2,
  hide.ns = FALSE,
  tip.length = 0)+
labs(
  caption = get_pwc_label(pair2)
)
```



Mixed Factorial ANOVA | APA Style ANOVA Tables

```
res.aov <- dat_MF %>% ezANOVA(dv = .(Time),
                             wid = .(Subject),
                             between= Condition,
                             within = .(Rank),
                             detailed = TRUE)

res.aov$ANOVA$ges <- res.aov$ANOVA$SSn/(res.aov$ANOVA$SSn+res.aov$ANOVA$SSd)

res.aov$ANOVA<-res.aov$ANOVA[c(1:3,6:7,9)]

apa.ezANOVA.table(res.aov,
                  table.title = "Mixed Factorial ANOVA",
                  correction="GG",
                  filename = "This is my file.doc",
                  table.number=7)
```

```
##
##
## Table 7
##
## Mixed Factorial ANOVA
##
##      Predictor df_num df_den Epsilon      F      p ges
##      (Intercept)   1.00  28.00      6339.90 .000 1.00
##      Condition     1.00  28.00       0.05 .825 .00
##      Rank          1.52  42.57    0.76 158.38 .000 .85
## Condition x Rank   1.52  42.57    0.76  10.57 .001 .27
##
## Note. df_num indicates degrees of freedom numerator. df_den indicates degrees of freedom denominator.
## Epsilon indicates Greenhouse-Geisser multiplier for degrees of freedom,
## p-values and degrees of freedom in the table incorporate this correction.
## ges indicates generalized eta-squared.
##
```

#Generate Marginal Means Table

```
apa.2way.table(Rank,
               Condition,
               Time,
               dat_MF,
               table.number=8,
               show.conf.interval = TRUE)
```

```
##
##
## Table 8
##
## Means and standard deviations for Time as a function of a 3(Rank) X 2(Condition) design
##
##           M           M_95%_CI    SD
## Condition:Low-Stress
##           Rank
##           Sophomore 503.27 [480.65, 525.88] 40.84
##           Junior 456.40 [430.56, 482.24] 46.66
##           Senior 412.67 [393.96, 431.37] 33.77
##
## Condition:High-Stress
##           Rank
##           Sophomore 534.40 [514.39, 554.41] 36.13
##           Junior 464.13 [449.23, 479.04] 26.91
##           Senior 381.53 [358.24, 404.82] 42.05
##
## Note. M and SD represent mean and standard deviation, respectively.
## LL and UL indicate the lower and upper limits of the
## 95% confidence interval for the mean, respectively.
## The confidence interval is a plausible range of population means
## that could have created a sample mean (Cumming, 2014).
```