

# STAT-S632

## Assignment 6

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```
import::from(magrittr, `>`, `<>`)
library(ggplot2)
import::from(lme4, lmer, VarCorr, ranef)
import::from(RLRsim, exactRLRT)

theme_set(theme_bw())
```

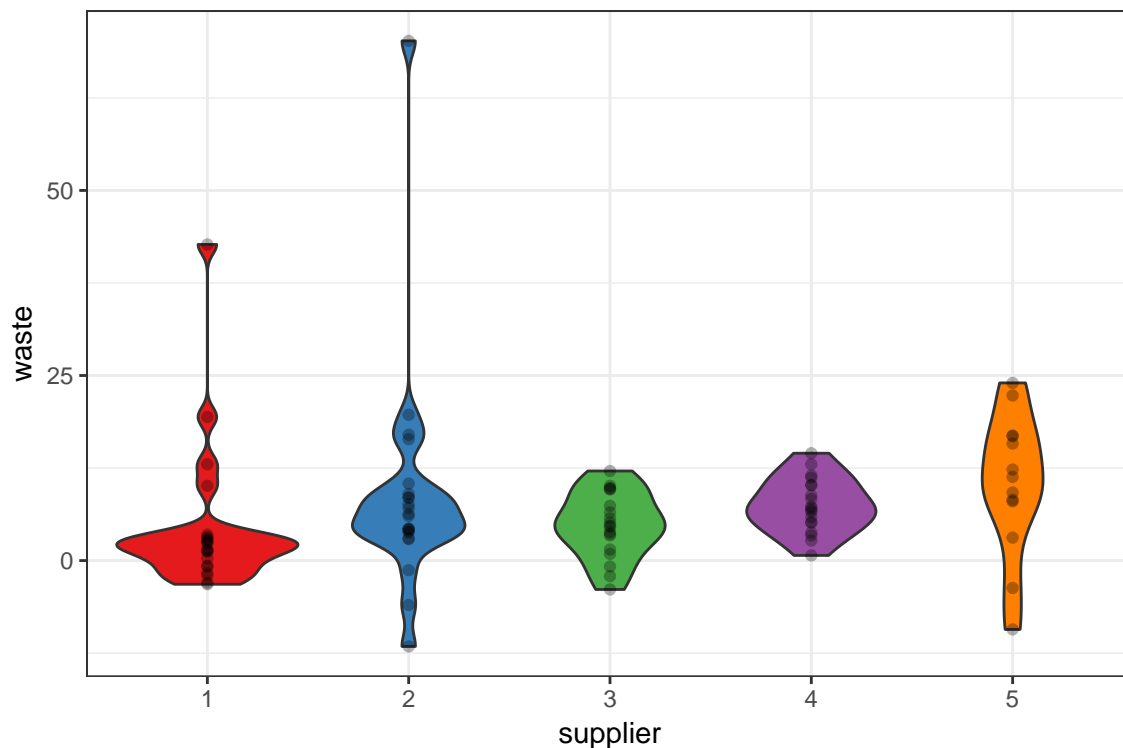
## Problem 1

```
denim.df <- faraway::denim
summary(denim.df)
```

	waste	supplier
Min.	:-11.600	1:22
1st Qu.:	2.550	2:22
Median :	5.200	3:19
Mean :	6.977	4:19
3rd Qu.:	9.950	5:13
Max.	: 70.200	

## Part a

```
ggplot(denim.df) +
  geom_violin(aes(x = supplier, y = waste, fill = supplier)) +
  geom_point(aes(x = supplier, y = waste), alpha = .3) +
  scale_fill_brewer(palette = 'Set1') +
  guides(fill = FALSE)
```



## Part b

```
fixed.mod <- lm(waste ~ supplier, data = denim.df)
summary(fixed.mod)
```

Call:

```
lm(formula = waste ~ supplier, data = denim.df)
```

Residuals:

Min	1Q	Median	3Q	Max
-20.432	-4.377	-1.323	2.639	61.368

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.5227	2.1021	2.152	0.0341 *
supplier2	4.3091	2.9728	1.450	0.1507
supplier3	0.3089	3.0879	0.100	0.9206
supplier4	2.9667	3.0879	0.961	0.3392
supplier5	5.8542	3.4491	1.697	0.0931 .

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 9.86 on 90 degrees of freedom

Multiple R-squared: 0.04901, Adjusted R-squared: 0.006747

F-statistic: 1.16 on 4 and 90 DF, p-value: 0.334

```
drop1(fixed.mod, test = 'F')
```

Single term deletions

Model:

waste ~ supplier

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)
<none>			8749.1	439.67		
supplier	4	450.92	9200.0	436.44	1.1596	0.334

supplier is not significant.

## Part c

```
rand.eff.mod <- lmer(waste ~ 1 + (1 | supplier), data = denim.df)
summary(rand.eff.mod)
```

Linear mixed model fit by REML ['lmerMod']

Formula: waste ~ 1 + (1 | supplier)

Data: denim.df

REML criterion at convergence: 702.1

Scaled residuals:

Min	1Q	Median	3Q	Max
-1.9095	-0.4363	-0.1669	0.3142	6.3817

Random effects:

Groups	Name	Variance	Std.Dev.
supplier	(Intercept)	0.6711	0.8192
	Residual	97.3350	9.8658

Number of obs: 95, groups: supplier, 5

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	6.997	1.078	6.49

$\beta = \mu \in \mathbb{R}^1$

$X \in \mathbb{R}^{95}$  (`nrow(denim.df)`)

$y, \epsilon \in \mathbb{R}^{95}$  as well

$\gamma \in \mathbb{R}^5$ , or the number of levels of `supplier`

$Z = \begin{bmatrix} z_1 & & & & \\ & z_2 & & & \\ & & z_3 & & \\ & & & z_4 & \\ & & & & z_5 \end{bmatrix}$  where  $\dim(z_i) = |\{x : x = \alpha_i\}|$ , or the number of elements in level  $i$ . So

$Z \in \mathbb{R}^{95 \times 5}$ .

## Part d

### Parametric bootstrapping

```
confint(rand.eff.mod, method = 'boot')
```

```
          2.5 %    97.5 %  
.sig01      0.000000  3.788285  
.sigma      8.325332 11.300242  
(Intercept) 4.866398  8.751813
```

Since the lower bound of the interval is 0, we cannot say that the variance of `supplier` is significant. We can also see this from the  $p$ -value.

```
exactRLRT(rand.eff.mod)
```

```
simulated finite sample distribution of RLRT.
```

```
(p-value based on 10000 simulated values)
```

```
data:
```

```
RLRT = 0.029383, p-value = 0.3491
```

### LR test

```
mle.mod <- lmer(waste ~ 1 + (1 | supplier), data = denim.df, REML = FALSE)  
pchisq(2 * (logLik(mle.mod) - logLik(fixed.mod)), 1, lower.tail = FALSE)
```

```
'log Lik.' 1 (df=3)
```

We get a  $p$ -value of 1.

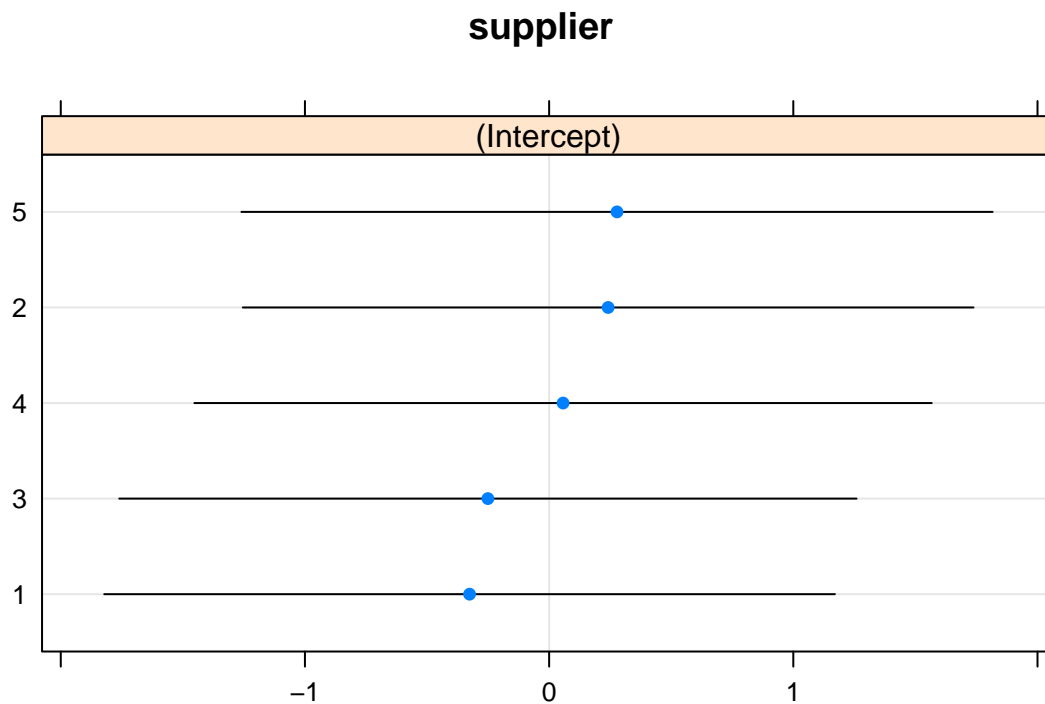
## Part e

```
ranef(rand.eff.mod)
```

```
$supplier  
(Intercept)  
1 -0.32586973  
2  0.24163762  
3 -0.25080816  
4  0.05703177  
5  0.27800850
```

```
lattice::dotplot(ranef(rand.eff.mod, condVar = TRUE))
```

```
$supplier
```



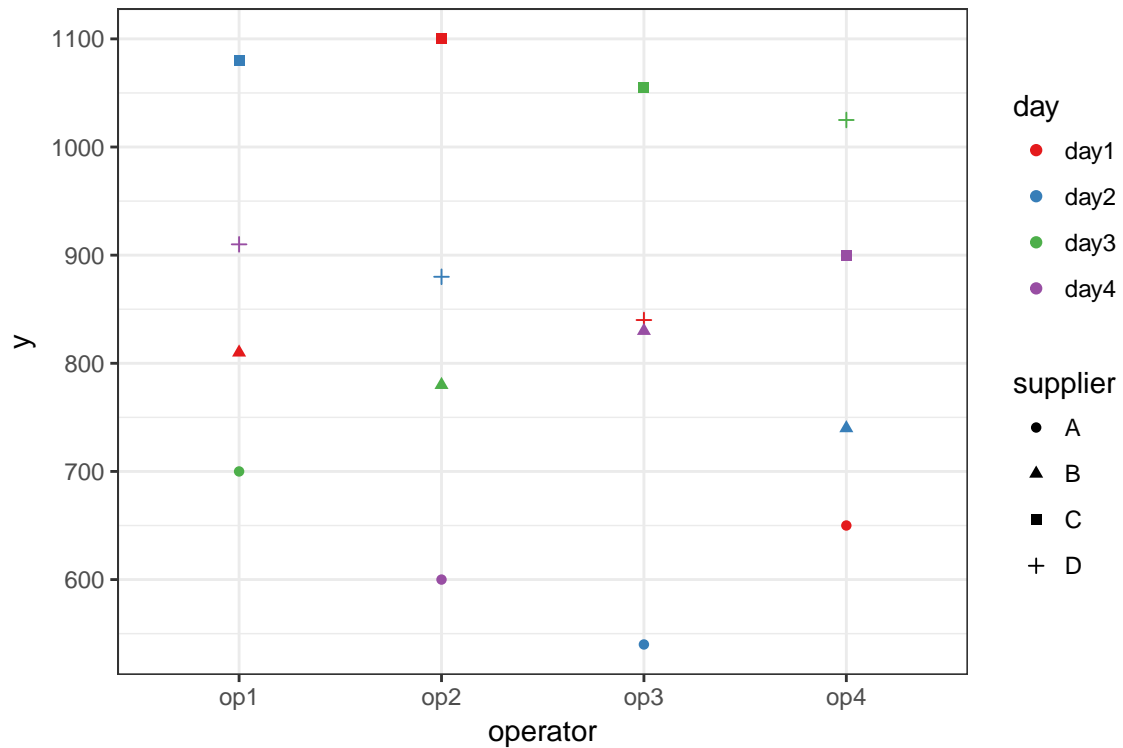
## Problem 2

```
breaking.df <- faraway::breaking
summary(breaking.df)
```

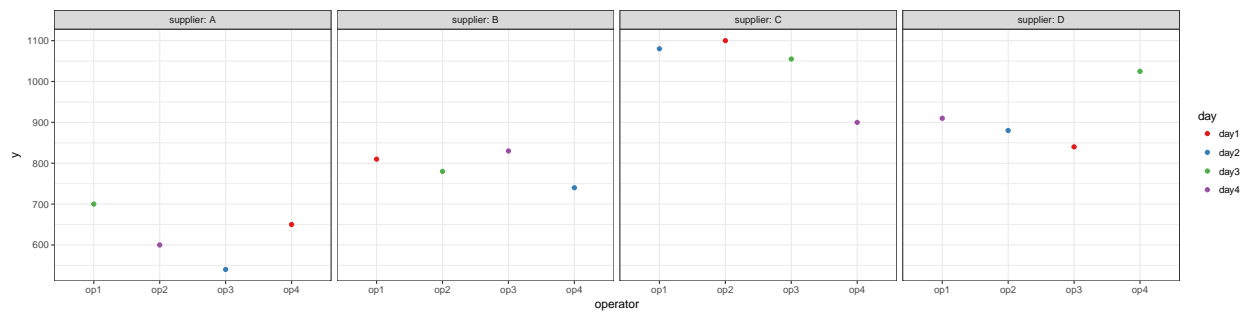
y	operator	day	supplier
Min. : 540.0	op1:4	day1:4	A:4
1st Qu.: 730.0	op2:4	day2:4	B:4
Median : 835.0	op3:4	day3:4	C:4
Mean : 840.0	op4:4	day4:4	D:4
3rd Qu.: 938.8			
Max. : 1100.0			

### Part a

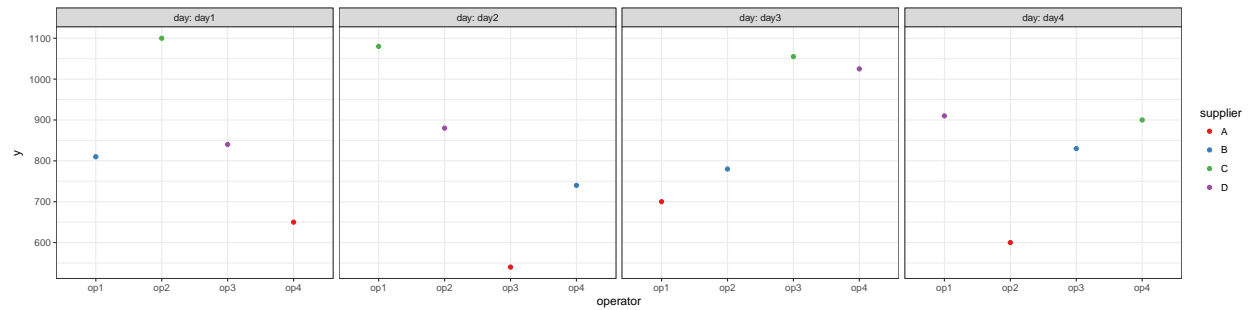
```
ggplot(breaking.df,
       aes(y = y, x = operator, colour = day, shape = supplier)) +
  geom_point() +
  scale_colour_brewer(palette = 'Set1')
```



```
ggplot(breaking.df,
  aes(y = y, x = operator, colour = day)) +
  geom_point() +
  scale_colour_brewer(palette = 'Set1') +
  facet_wrap(~ supplier, labeller = 'label_both', nrow = 1)
```



```
ggplot(breaking.df,
  aes(y = y, x = operator, colour = supplier)) +
  geom_point() +
  scale_colour_brewer(palette = 'Set1') +
  facet_wrap(~ day, labeller = 'label_both', nrow = 1)
```



There appears to be a suggestion of differences among suppliers but not among operators or days.

## Part b

$X \in \mathbb{R}^{16 \times 4}$  (16 rows  $\times$  4 factor levels)

$\beta \in \mathbb{R}^4$  (4 factor levels)

$y, \epsilon \in \mathbb{R}^{16}$  (16 rows)

$\gamma \in \mathbb{R}^7$  (1 intercept + (3 - 1) from factor 1 + (3 - 1) from factor 2)

$Z \in \mathbb{R}^{16 \times 7}$ , or the number of rows  $\times$  dim( $\gamma$ )

## Part c

```
fixed.mod <- lm(y ~ day + operator + supplier, data = breaking.df)
summary(fixed.mod)
```

Call:

```
lm(formula = y ~ day + operator + supplier, data = breaking.df)
```

Residuals:

Min	1Q	Median	3Q	Max
-92.50	-25.94	-6.25	31.88	93.75

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	667.50	62.29	10.716	3.9e-05 ***
dayday2	-40.00	55.72	-0.718	0.499782
dayday3	40.00	55.72	0.718	0.499782
dayday4	-40.00	55.72	-0.718	0.499782
operatorop2	-35.00	55.72	-0.628	0.553020
operatorop3	-58.75	55.72	-1.054	0.332266
operatorop4	-46.25	55.72	-0.830	0.438247
supplierB	167.50	55.72	3.006	0.023812 *
supplierC	411.25	55.72	7.381	0.000317 ***
supplierD	291.25	55.72	5.227	0.001962 **

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 78.79 on 6 degrees of freedom

Multiple R-squared: 0.9141, Adjusted R-squared: 0.7853  
F-statistic: 7.094 on 9 and 6 DF, p-value: 0.01348

```
drop1(fixed.mod, test = 'F')
```

Single term deletions

Model:

```
y ~ day + operator + supplier
```

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)
<none>			37250	144.04		
day	3	17600	54850	144.24	0.9450	0.475896
operator	3	7662	44912	141.04	0.4114	0.750967
supplier	3	371137	408387	176.36	19.9268	0.001602 **

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
anova(fixed.mod, lm(y ~ supplier, data = breaking.df))
```

Analysis of Variance Table

Model 1: y ~ day + operator + supplier

Model 2: y ~ supplier

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	6	37250				
2	12	62513	-6	-25262	0.6782	0.6754

```
fixed.mod <- lm(y ~ supplier, data = breaking.df)  
summary(fixed.mod)
```

Call:

```
lm(formula = y ~ supplier, data = breaking.df)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-133.750	-37.813	8.125	41.563	111.250

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	622.50	36.09	17.250	7.79e-10 ***
supplierB	167.50	51.04	3.282	0.00656 **
supplierC	411.25	51.04	8.058	3.49e-06 ***
supplierD	291.25	51.04	5.707	9.81e-05 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 72.18 on 12 degrees of freedom

Multiple R-squared: 0.8558, Adjusted R-squared: 0.8198

F-statistic: 23.75 on 3 and 12 DF, p-value: 2.464e-05

It appears that supplier is the only significant covariate.



## Part d

```
mixed.mod <- lmer(y ~ supplier + (1 | operator) + (1 | day), data = breaking.df)
summary(mixed.mod)
```

```
Linear mixed model fit by REML ['lmerMod']
Formula: y ~ supplier + (1 | operator) + (1 | day)
Data: breaking.df
```

REML criterion at convergence: 142.3

Scaled residuals:

Min	1Q	Median	3Q	Max
-1.8299	-0.4718	0.1027	0.6518	1.4691

Random effects:

Groups	Name	Variance	Std.Dev.
operator	(Intercept)	7.275e-13	8.529e-07
day	(Intercept)	2.191e+02	1.480e+01
Residual		4.990e+03	7.064e+01

Number of obs: 16, groups: operator, 4; day, 4

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	622.50	36.09	17.250
supplierB	167.50	49.95	3.353
supplierC	411.25	49.95	8.233
supplierD	291.25	49.95	5.831

Correlation of Fixed Effects:

	(Intr)	spplrB	spplrC
supplierB	-0.692		
supplierC	-0.692	0.500	
supplierD	-0.692	0.500	0.500

The experiment was conducted to select a supplier, so it is the effect of interest.

Supplier C has the highest breaking point, as we saw in the fixed effect model and the plots.

## Part e

```
mixed.op.mod <- lmer(y ~ supplier + (1 | operator), data = breaking.df)
mixed.day.mod <- lmer(y ~ supplier + (1 | day), data = breaking.df)

exactRLRT(mixed.op.mod, mixed.mod, mixed.day.mod)
```

simulated finite sample distribution of RLRT.

(p-value based on 10000 simulated values)

data:

RLRT = 0, p-value = 1

```
exactRLRT(mixed.day.mod, mixed.mod, mixed.op.mod)
```

simulated finite sample distribution of RLRT.

(p-value based on 10000 simulated values)

data:

RLRT = 0.030235, p-value = 0.3716

Neither are significant.