

Homework 12

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6210422036: Tanat Iempreedee

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```
library(dplyr)
library(ggplot2)
library(caret)
library(e1071)
library(reshape2)

d <- read.table('shipment.csv', sep=',', header=TRUE,
               colClasses=c('factor','numeric','numeric','factor'))

str(d)
```

```
'data.frame':  400 obs. of  4 variables:
 $ Late      : Factor w/ 2 levels "0","1": 1 2 2 2 1 2 2 1 2 1 ...
 $ Distance: num  380 660 800 640 520 760 560 400 540 700 ...
 $ Weight   : num  3.61 3.67 4 3.19 2.93 3 2.98 3.08 3.39 3.92 ...
 $ LSPRank  : Factor w/ 4 levels "1","2","3","4": 3 3 1 4 4 2 1 2 3 2 ...
```

1) EDA

Among all late shipments, the average distance and weight are higher than on-time shipments.

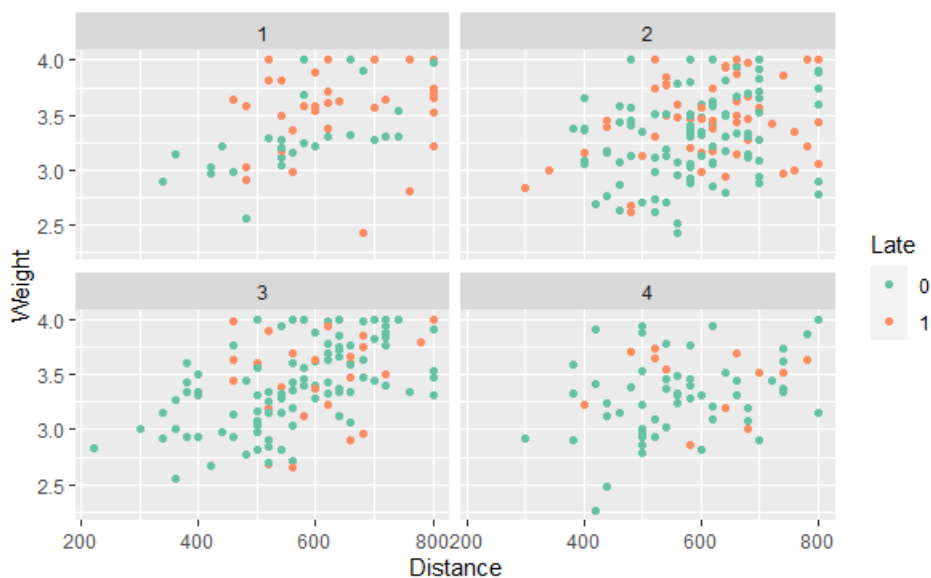
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```
d %>% group_by(Late) %>%
  summarise(AvgDistance=mean(Distance), AvgWeight=mean(Weight), n=n())
```

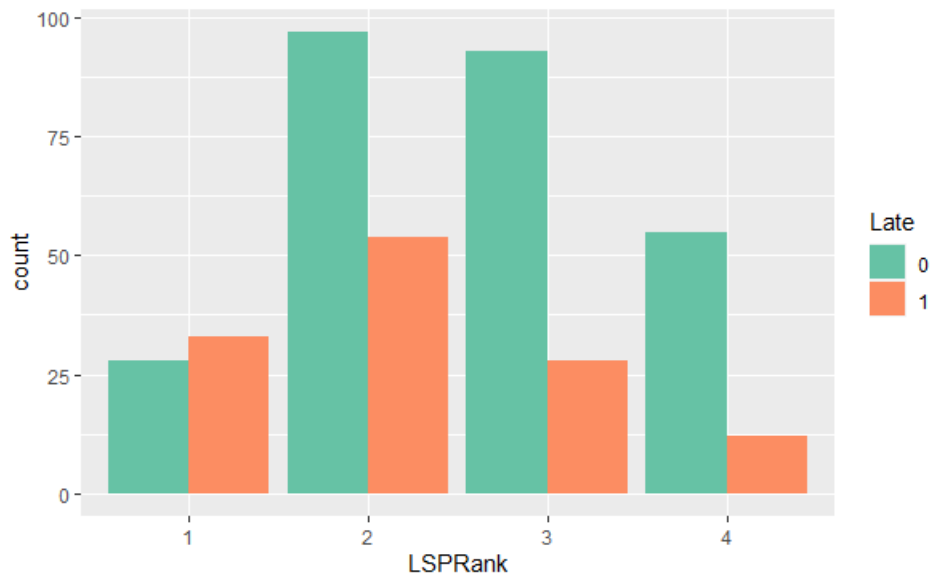
Late <fctr>	AvgDistance <dbl>	AvgWeight <dbl>	n <int>
0	573.1868	3.343700	273
1	618.8976	3.489213	127
2 rows			

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```
# EDA
ggplot(d, aes(Distance, Weight)) +
  geom_point(aes(color=Late)) +
  facet_wrap(~LSPRank) +
  scale_color_brewer(palette="Set2")
```

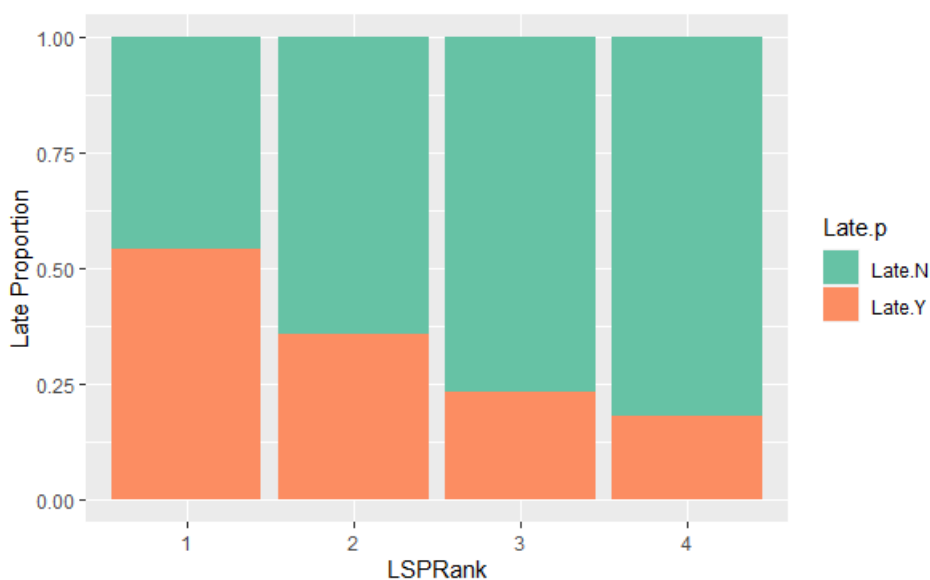


```
# count plot
ggplot(d, aes(fill=Late, x=LSPRank)) +
  geom_bar(position="dodge") +
  scale_fill_brewer(palette="Set2")
```



```
# normalized count plot
d %>% group_by(LSPRank) %>%
  summarise(Late.N=mean(Late=='0'), Late.Y=mean(Late=='1')) %>%
  melt(variable.name='Late.p') %>%
  ggplot(aes(x=LSPRank, y=value, fill=Late.p)) +
  geom_bar(position='stack', stat='identity') +
  ylab('Late Proportion') +
  scale_fill_brewer(palette="Set2")
```

Using LSPRank as id variables



2) Fit Logistic Regression

```
logreg <- glm(Late~Distance + Weight + LSPRank, data=d, family="binomial")
summary(logreg)
```

```
Call:
glm(formula = Late ~ Distance + Weight + LSPRank, family = "binomial",
    data = d)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.6268	-0.8662	-0.6388	1.1490	2.0790

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-3.989979	1.139951	-3.500	0.000465	***
Distance	0.002264	0.001094	2.070	0.038465	*
Weight	0.804038	0.331819	2.423	0.015388	*
LSPRank2	-0.675443	0.316490	-2.134	0.032829	*
LSPRank3	-1.340204	0.345306	-3.881	0.000104	***
LSPRank4	-1.551464	0.417832	-3.713	0.000205	***

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

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 499.98 on 399 degrees of freedom
Residual deviance: 458.52 on 394 degrees of freedom
AIC: 470.52

Number of Fisher Scoring iterations: 4

Coefficient interpretation

- Increases in **Distance** and **Weight** raises the probability of being late. and Weight has a higher effect. -> make sense
- LSPRank** of 2,3,4 has less probability of being Late compared to LSPRank 1. The higher ranks of LSPRank decreases more probability of being late. This seems to make sense if higher rank means higher quality of a service provider.

3) Predict

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```
newdata <- data.frame(Distance=680,Weight=3.7,LSPRank=factor(2, levels=1:4))
newdata
```

Distance	Weight	LSPRank
<dbl>	<dbl>	<fctr>
680	3.7	2

1 row

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```
y_pred <- predict(logreg, newdata=newdata ,type='response')
y_pred
```

```
1
0.4624027
```

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```
ifelse(y_pred >= 0.5, 'Late', 'Not Late')
```

```
1
"Not Late"
```