

HW3

2024-11-25

#Preparation

```
# 載入資料
library(readr)
data <- read_csv("C:/Users/Ava/Desktop/R/HW3/airline_survey.csv")
```

```
## New names:
## Rows: 103904 Columns: 25
## — Column specification
## _____ Delimiter: "," chr
## (5): Gender, Customer Type, Type of Travel, Class, satisfaction dbl (20): ...1,
## id, Age, Flight Distance, Inflight wifi service, Departure/A...
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## • `` -> `...1`
```

```
View(data)
```

```
# summary
#install.packages("summarytools")
library(summarytools)
```

```
## Warning: 套件 'summarytools' 是用 R 版本 4.4.2 來建造的
```

```
dfSummary(data)
```

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```

IIIIIIIIIIII 103904 0
## [character] 2. Personal Travel 32249 (31.0%)
IIIIII (100.0%) (0.0%)
##
## 7 Class 1. Business 49665 (47.8%)
IIIIIIII 103904 0
## [character] 2. Eco 46745 (45.0%)
IIIIIIII (100.0%) (0.0%)
## 3. Eco Plus 7494 ( 7.2%)
I
##
## 8 Flight Distance Mean (sd) : 1189.4 (997.1) 3802 distinct val
ues : 103904 0
## [numeric] min < med < max:
: : (100.0%) (0.0%)
## 31 < 843 < 4983
: :
## IQR (CV) : 1329 (0.8)
: : . .
##
## 9 Inflight wifi service Mean (sd) : 2.7(1.3) 0 : 3103 ( 3.0%)
103904 0
## [numeric] min < med < max: 1 : 17840 (17.2%)
III (100.0%) (0.0%) 0 < 3 < 5 2 : 25830 (24.9%)
## IQR (CV) : 2 (0.5) 3 : 25868 (24.9%)
IIII 4 : 19794 (19.1%)
## 5 : 11469 (11.0%)
III
##
II
##
## 10 Departure/Arrival time convenient Mean (sd) : 3.1 (1.5) 0 : 5300 ( 5.1%)
I 103904 0
## [numeric] min < med < max: 1 : 15498 (14.9%)
II (100.0%) (0.0%) 0 < 3 < 5 2 : 17191 (16.5%)
## IQR (CV) : 2 (0.5) 3 : 17966 (17.3%)
III 4 : 25546 (24.6%)
## 5 : 22403 (21.6%)
IIII
##
## 11 Ease of Online booking Mean (sd) : 2.8 (1.4) 0 : 4487 ( 4.3%)
103904 0
## [numeric] min < med < max: 1 : 17525 (16.9%)
III (100.0%) (0.0%) 0 < 3 < 5 2 : 24021 (23.1%)
## IQR (CV) : 2 (0.5) 3 : 24449 (23.5%)
IIII

```

##				4 : 19571 (18.8%)
III				
##				5 : 13851 (13.3%)
II				
##				
## 12	Gate location	Mean (sd) : 3 (1.3)	0 :	1 (0.0%)
103904	0			
##	[numeric]	min < med < max:	1 :	17562 (16.9%)
III	(100.0%) (0.0%)		2 :	19459 (18.7%)
##		0 < 3 < 5	3 :	28577 (27.5%)
III			4 :	24426 (23.5%)
##		IQR (CV) : 2 (0.4)	5 :	13879 (13.4%)
II				
##				
## 13	Food and drink	Mean (sd) : 3.2 (1.3)	0 :	107 (0.1%)
103904	0			
##	[numeric]	min < med < max:	1 :	12837 (12.4%)
II	(100.0%) (0.0%)		2 :	21988 (21.2%)
##		0 < 3 < 5	3 :	22300 (21.5%)
IIII			4 :	24359 (23.4%)
##		IQR (CV) : 2 (0.4)	5 :	22313 (21.5%)
IIII				
##				
## 14	Online boarding	Mean (sd) : 3.3 (1.3)	0 :	2428 (2.3%)
103904	0			
##	[numeric]	min < med < max:	1 :	10692 (10.3%)
II	(100.0%) (0.0%)		2 :	17505 (16.8%)
##		0 < 3 < 5	3 :	21804 (21.0%)
III			4 :	30762 (29.6%)
##		IQR (CV) : 2 (0.4)	5 :	20713 (19.9%)
IIII				
##				
## 15	Seat comfort	Mean (sd) : 3.4 (1.3)	0 :	1 (0.0%)
103904	0			
##	[numeric]	min < med < max:	1 :	12075 (11.6%)
II	(100.0%) (0.0%)		2 :	14897 (14.3%)
##		0 < 4 < 5	3 :	18696 (18.0%)
II			4 :	31765 (30.6%)
##		IQR (CV) : 3 (0.4)	5 :	26470 (25.5%)
III				
##				
IIIIII				
##				
IIII				

##					
## 16	Inflight entertainment		Mean (sd) : 3.4 (1.3)	0 :	14 (0.0%)
103904	0				
##	[numeric]		min < med < max:	1 :	12478 (12.0%)
II	(100.0%)	(0.0%)			
##			0 < 4 < 5	2 :	17637 (17.0%)
III					
##			IQR (CV) : 2 (0.4)	3 :	19139 (18.4%)
III					
##				4 :	29423 (28.3%)
IIIII					
##				5 :	25213 (24.3%)
IIII					
##					
## 17	On-board service		Mean (sd) : 3.4 (1.3)	0 :	3 (0.0%)
103904	0				
##	[numeric]		min < med < max:	1 :	11872 (11.4%)
II	(100.0%)	(0.0%)			
##			0 < 4 < 5	2 :	14681 (14.1%)
II					
##			IQR (CV) : 2 (0.4)	3 :	22833 (22.0%)
IIII					
##				4 :	30867 (29.7%)
IIIII					
##				5 :	23648 (22.8%)
IIII					
##					
## 18	Leg room service		Mean (sd) : 3.4 (1.3)	0 :	472 (0.5%)
103904	0				
##	[numeric]		min < med < max:	1 :	10353 (10.0%)
I	(100.0%)	(0.0%)			
##			0 < 4 < 5	2 :	19525 (18.8%)
III					
##			IQR (CV) : 2 (0.4)	3 :	20098 (19.3%)
III					
##				4 :	28789 (27.7%)
IIIII					
##				5 :	24667 (23.7%)
IIII					
##					
## 19	Baggage handling		Mean (sd) : 3.6 (1.2)	1 :	7237 (7.0%)
I	103904	0			
##	[numeric]		min < med < max:	2 :	11521 (11.1%)
II	(100.0%)	(0.0%)			
##			1 < 4 < 5	3 :	20632 (19.9%)
III					
##			IQR (CV) : 2 (0.3)	4 :	37383 (36.0%)
IIIIIIII					
##				5 :	27131 (26.1%)
IIIII					
##					
## 20	Checkin service		Mean (sd) : 3.3 (1.3)	0 :	1 (0.0%)
103904	0				
##	[numeric]		min < med < max:	1 :	12890 (12.4%)
II	(100.0%)	(0.0%)			
##			0 < 3 < 5	2 :	12893 (12.4%)

```

II
##                                IQR (CV) : 1 (0.4)                                3 : 28446 (27.4%)
IIIII
##                                4 : 29055 (28.0%)
IIIII
##                                5 : 20619 (19.8%)
III
##
## 21 Inflight service                Mean (sd) : 3.6 (1.2)                0 :      3 ( 0.0%)
103904      0
##      [numeric]                min < med < max:                1 :  7084 ( 6.8%)
I                                (100.0%)  (0.0%)
##                                0 < 4 < 5                2 : 11457 (11.0%)
II
##                                IQR (CV) : 2 (0.3)                3 : 20299 (19.5%)
III
##                                4 : 37945 (36.5%)
IIIIIII
##                                5 : 27116 (26.1%)
IIIII
##
## 22 Cleanliness                Mean (sd) : 3.3 (1.3)                0 :     12 ( 0.0%)
103904      0
##      [numeric]                min < med < max:                1 : 13318 (12.8%)
II                                (100.0%)  (0.0%)
##                                0 < 3 < 5                2 : 16132 (15.5%)
III
##                                IQR (CV) : 2 (0.4)                3 : 24574 (23.7%)
IIII
##                                4 : 27179 (26.2%)
IIIII
##                                5 : 22689 (21.8%)
IIII
##
## 23 Departure Delay in Minutes    Mean (sd) : 14.8 (38.2)            446 distinct valu
es      :                        103904      0
##      [numeric]                min < med < max:
:                                (100.0%)  (0.0%)
##                                0 < 0 < 1592
:
##                                IQR (CV) : 12 (2.6)
:
##
:
##
## 24 Arrival Delay in Minutes      Mean (sd) : 15.2 (38.7)            455 distinct valu
es      :                        103594      310
##      [numeric]                min < med < max:
:                                (99.7%)  (0.3%)
##                                0 < 0 < 1584
:
##                                IQR (CV) : 13 (2.5)
:
##
:
##

```

```
## 25    satisfaction                                1. neutral or dissatisfied    58879 (56.7%)
IIIIIIIIII      103904      0
##      [character]                                2. satisfied                45025 (43.3%)
IIIIIIII      (100.0%)    (0.0%)
## -----
-----
```

```
view(dfSummary(data)) # 看資料的各項分布
```

```
## Switching method to 'browser'
```

```
## Output file written: C:\Users\Ava\AppData\Local\Temp\Rtmp4Q0Bm6\file2a5c745638dc.html
```

```
freq(data)
```

```
## Variable(s) ignored: ...1, id, Age, Flight Distance, Departure Delay in Minutes, Arrival Delay in Minutes
```

Frequencies

data\$Gender

Type: Character

##

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
Female	52727	50.75	50.75	50.75	50.75
Male	51177	49.25	100.00	49.25	100.00
<NA>	0			0.00	100.00
Total	103904	100.00	100.00	100.00	100.00

##

data\$Customer Type

Type: Character

##

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
disloyal Customer	18981	18.27	18.27	18.27	18.27
Loyal Customer	84923	81.73	100.00	81.73	100.00
<NA>	0			0.00	100.00
Total	103904	100.00	100.00	100.00	100.00

##

data\$Type of Travel

Type: Character

##

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
Business travel	71655	68.96	68.96	68.96	68.96
Personal Travel	32249	31.04	100.00	31.04	100.00
<NA>	0			0.00	100.00
Total	103904	100.00	100.00	100.00	100.00

##

data\$Class

Type: Character

##

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
Business	49665	47.80	47.80	47.80	47.80
Eco	46745	44.99	92.79	44.99	92.79
Eco Plus	7494	7.21	100.00	7.21	100.00
<NA>	0			0.00	100.00
Total	103904	100.00	100.00	100.00	100.00

##

data\$Inflight wifi service

Type: Numeric

##

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
0	3103	2.99	2.99	2.99	2.99
1	17840	17.17	20.16	17.17	20.16
2	25830	24.86	45.02	24.86	45.02
3	25868	24.90	69.91	24.90	69.91
4	19794	19.05	88.96	19.05	88.96
5	11469	11.04	100.00	11.04	100.00
<NA>	0			0.00	100.00
Total	103904	100.00	100.00	100.00	100.00


```
##
## data$Departure/Arrival time convenient
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0    5300      5.10         5.10     5.10         5.10
##          1   15498     14.92        20.02    14.92        20.02
##          2   17191     16.55        36.56    16.55        36.56
##          3   17966     17.29        53.85    17.29        53.85
##          4   25546     24.59        78.44    24.59        78.44
##          5   22403     21.56       100.00    21.56       100.00
##         <NA>         0              0.00    0.00       100.00
##        Total 103904    100.00       100.00  100.00       100.00
##
## data$Ease of Online booking
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0    4487      4.32         4.32     4.32         4.32
##          1   17525     16.87        21.18    16.87        21.18
##          2   24021     23.12        44.30    23.12        44.30
##          3   24449     23.53        67.83    23.53        67.83
##          4   19571     18.84        86.67    18.84        86.67
##          5   13851     13.33       100.00    13.33       100.00
##         <NA>         0              0.00    0.00       100.00
##        Total 103904    100.00       100.00  100.00       100.00
##
## data$Gate location
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0         1    0.00096    0.00096    0.00096    0.00096
##          1   17562    16.90214    16.90310    16.90214    16.90310
##          2   19459    18.72786    35.63097    18.72786    35.63097
##          3   28577    27.50327    63.13424    27.50327    63.13424
##          4   24426    23.50824    86.64248    23.50824    86.64248
##          5   13879    13.35752   100.00000    13.35752   100.00000
##         <NA>         0              0.00000    0.00000   100.00000
##        Total 103904   100.00000   100.00000  100.00000   100.00000
##
## data$Food and drink
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0     107      0.10         0.10     0.10         0.10
##          1   12837     12.35        12.46    12.35        12.46
##          2   21988     21.16        33.62    21.16        33.62
##          3   22300     21.46        55.08    21.46        55.08
##          4   24359     23.44        78.53    23.44        78.53
##          5   22313     21.47       100.00    21.47       100.00
##         <NA>         0              0.00    0.00       100.00
##        Total 103904    100.00       100.00  100.00       100.00
```

```
##
## data$Online boarding
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0    2428     2.34         2.34     2.34         2.34
##          1   10692    10.29        12.63    10.29        12.63
##          2   17505    16.85        29.47    16.85        29.47
##          3   21804    20.98        50.46    20.98        50.46
##          4   30762    29.61        80.07    29.61        80.07
##          5   20713    19.93       100.00    19.93       100.00
##         <NA>         0              0.00    100.00
##        Total 103904   100.00       100.00   100.00       100.00
##
## data$Seat comfort
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0         1    0.00096     0.00096    0.00096     0.00096
##          1   12075   11.62130    11.62227   11.62130    11.62227
##          2   14897   14.33727    25.95954   14.33727    25.95954
##          3   18696   17.99353    43.95307   17.99353    43.95307
##          4   31765   30.57149    74.52456   30.57149    74.52456
##          5   26470   25.47544   100.00000   25.47544   100.00000
##         <NA>         0              0.00000   100.00000
##        Total 103904  100.00000   100.00000  100.00000   100.00000
##
## data$Inflight entertainment
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0        14     0.013     0.013     0.013     0.013
##          1   12478   12.009     12.023    12.009     12.023
##          2   17637   16.974     28.997    16.974     28.997
##          3   19139   18.420     47.417    18.420     47.417
##          4   29423   28.317     75.734    28.317     75.734
##          5   25213   24.266    100.000    24.266    100.000
##         <NA>         0              0.000     100.000
##        Total 103904  100.000     100.000   100.000     100.000
##
## data$On-board service
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0         3    0.0029     0.0029    0.0029     0.0029
##          1   11872   11.4259    11.4288   11.4259    11.4288
##          2   14681   14.1294    25.5582   14.1294    25.5582
##          3   22833   21.9751    47.5333   21.9751    47.5333
##          4   30867   29.7072    77.2405   29.7072    77.2405
##          5   23648   22.7595   100.0000   22.7595   100.0000
##         <NA>         0              0.0000   100.0000
##        Total 103904  100.0000   100.0000  100.0000   100.0000
```

```
##
## data$Leg room service
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0    472      0.45          0.45    0.45      0.45
##          1   10353     9.96         10.42    9.96     10.42
##          2   19525    18.79         29.21   18.79     29.21
##          3   20098    19.34         48.55   19.34     48.55
##          4   28789    27.71         76.26   27.71     76.26
##          5   24667    23.74        100.00   23.74    100.00
##         <NA>      0
##        Total 103904   100.00        100.00  100.00    100.00
##
## data$Baggage handling
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          1    7237     6.97          6.97    6.97     6.97
##          2   11521    11.09         18.05   11.09     18.05
##          3   20632    19.86         37.91   19.86     37.91
##          4   37383    35.98         73.89   35.98     73.89
##          5   27131    26.11        100.00   26.11    100.00
##         <NA>      0
##        Total 103904   100.00        100.00  100.00    100.00
##
## data$Checkin service
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0      1    0.00096    0.00096    0.00096    0.00096
##          1   12890   12.40568   12.40664   12.40568   12.40664
##          2   12893   12.40857   24.81521   12.40857   24.81521
##          3   28446   27.37719   52.19241   27.37719   52.19241
##          4   29055   27.96331   80.15572   27.96331   80.15572
##          5   20619   19.84428  100.00000   19.84428  100.00000
##         <NA>      0
##        Total 103904  100.00000  100.00000  100.00000  100.00000
##
## data$Inflight service
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0      3    0.0029    0.0029    0.0029    0.0029
##          1    7084    6.8178    6.8207    6.8178    6.8207
##          2   11457   11.0265   17.8472   11.0265   17.8472
##          3   20299   19.5363   37.3835   19.5363   37.3835
##          4   37945   36.5193   73.9028   36.5193   73.9028
##          5   27116   26.0972  100.0000   26.0972  100.0000
##         <NA>      0
##        Total 103904  100.0000  100.0000  100.0000  100.0000
##
```

```
## data$Cleanliness
## Type: Numeric
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
##          0      12    0.012      0.012    0.012      0.012
##          1    13318   12.818     12.829   12.818     12.829
##          2    16132   15.526     28.355   15.526     28.355
##          3    24574   23.651     52.006   23.651     52.006
##          4    27179   26.158     78.163   26.158     78.163
##          5    22689   21.837    100.000   21.837    100.000
##         <NA>        0          0.000   100.000
##        Total 103904  100.000    100.000  100.000    100.000
##
## data$satisfaction
## Type: Character
##
##          Freq  % Valid  % Valid Cum.  % Total  % Total Cum.
## -----
## neutral or dissatisfied  58879   56.67      56.67    56.67    56.67
##          satisfied      45025   43.33    100.00    43.33   100.00
##          <NA>           0          0.00    100.00
##          Total  103904  100.00    100.00   100.00   100.00
```

```
view(freq(data))
```

```
## Variable(s) ignored: ...1, id, Age, Flight Distance, Departure Delay in Minutes, Arrival Delay in Minutes
```

```
## Output file written: C:\Users\Ava\AppData\Local\Temp\Rtmp4Q0Bm6\file2a5c34f8454c.html
```

```
## Switching method to 'browser'
```

```
## Output file appended: C:\Users\Ava\AppData\Local\Temp\Rtmp4Q0Bm6\file2a5c34f8454c.html
```

```
# 資料清洗
```

```
missing_arrival_delay <- data[is.na(data$`Arrival Delay in Minutes`), ]
sum(is.na(data$`Arrival Delay in Minutes`)) # 確認有幾筆缺失值
```

```
## [1] 310
```

```
View(missing_arrival_delay) # 看有缺失值的筆數細項
data_cleaned <- na.omit(data) # 直接刪除有缺失值的資料
View(data_cleaned)
```

Q2.Customer segmentation

```
#install.packages("factoextra")
library(factoextra)
```

```
## Warning: 套件 'factoextra' 是用 R 版本 4.4.2 來建造的
```

```
## 載入需要的套件 : ggplot2
```

```
## Warning: 套件 'ggplot2' 是用 R 版本 4.4.2 來建造的
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
# 決定最佳群數(使用肘部法則)
# fviz_nbclust(data_cleaned[,9:22], FUNcluster = kmeans, method = "wss", k.max = 20) +
#   labs(title="Elbow Method for K-Means") +
#   geom_vline(xintercept = 3, linetype = 2)
# (我電腦沒有40GB可以跑這行程式碼，所以後面最佳群數用假設的)
```

```
# K-means 分群
k = kmeans(data_cleaned[,9:22], centers=3) # 選取服務滿意度的欄位分群
str(k)
```

```
## List of 9
## $ cluster      : int [1:103594] 3 1 3 1 2 1 1 2 1 1 ...
## $ centers      : num [1:3, 1:14] 2.31 3.9 1.91 3.02 3.98 ...
## .. attr(*, "dimnames")=List of 2
## .. ..$ : chr [1:3] "1" "2" "3"
## .. ..$ : chr [1:14] "Inflight wifi service" "Departure/Arrival time convenient" "Ease of
Online booking" "Gate location" ...
## $ totss       : num 2514603
## $ withinss    : num [1:3] 738628 508654 555576
## $ tot.withinss: num 1802858
## $ betweenss   : num 711745
## $ size        : int [1:3] 37396 35217 30981
## $ iter        : int 3
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

```
k$centers
```

```
## Inflight wifi service Departure/Arrival time convenient
## 1 2.305059 3.016285
## 2 3.900048 3.978703
## 3 1.912075 2.068720
## Ease of Online booking Gate location Food and drink Online boarding
## 1 2.587122 2.983929 2.163681 2.555006
## 2 3.889343 3.703439 3.716103 4.062640
## 3 1.674833 2.142959 3.871340 3.166812
## Seat comfort Inflight entertainment On-board service Leg room service
## 1 2.329233 2.003182 2.728019 2.846401
## 2 4.039867 4.086407 3.831048 3.768606
## 3 4.098092 4.166489 3.662987 3.486718
## Baggage handling Checkin service Inflight service Cleanliness
## 1 3.106803 2.876885 3.104637 2.106001
## 2 4.000227 3.612517 4.008888 3.888122
## 3 3.846325 3.469933 3.869436 4.027210
```

```
k$withinss
```

```
## [1] 738627.9 508654.1 555576.1
```

```
k$tot.withinss
```

```
## [1] 1802858
```

```
k$size
```

```
## [1] 37396 35217 30981
```

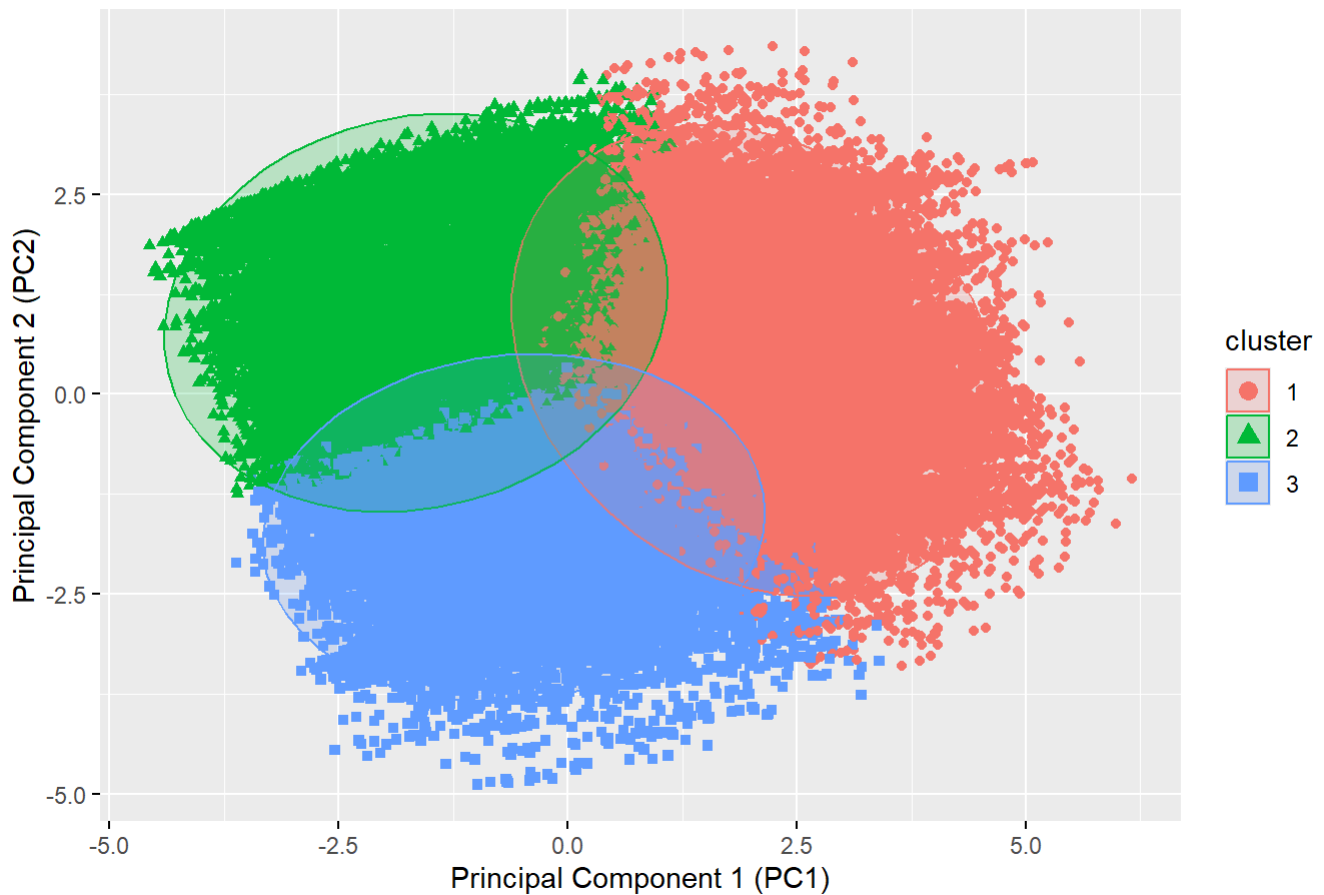
```
data_cleaned$Cluster <- k$cluster
```

```
# 視覺化分群結果
#install.packages("useful")
library(useful)
```

```
## Warning: 套件 'useful' 是用 R 版本 4.4.2 來建造的
```

```
fviz_cluster(k,
  data = data_cleaned[,9:22],
  geom = c("point"),
  ellipse.type = "norm") +
labs(title = "K-means Clustering",
  x = "Principal Component 1 (PC1)",
  y = "Principal Component 2 (PC2)")
```

K-means Clustering



```
# EDA
#install.packages("tidyverse")
library(tidyverse)
```

```
## Warning: 套件 'tidyverse' 是用 R 版本 4.4.2 來建造的
```

```
## — Attaching core tidyverse packages ————— tidyverse 2.0.0 —
## ✓ dplyr      1.1.4      ✓ stringr    1.5.1
## ✓ forcats    1.0.0      ✓ tibble     3.2.1
## ✓ lubridate  1.9.3      ✓ tidyr      1.3.1
## ✓ purrr      1.0.2
## — Conflicts ————— tidyverse_conflicts() —
## X dplyr::filter() masks stats::filter()
## X dplyr::lag()    masks stats::lag()
## X tibble::view()  masks summarytools::view()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to be
come errors
```

```
plot_satisfaction_proportion <- function(df, group) {
  service_cols <- names(df)[9:22]
  service_cols <- service_cols[apply(df[service_cols], is.numeric)]
  # 上面那行是用來確定每一個都是數值型態·理論上應該都是啦·但以防萬一檢查一下

  lowPer <- c()
  neuPer <- c()
  highPer <- c()

  for (col in service_cols) {
    rCount_low <- sum(df[[col]] %in% c(0, 1, 2), na.rm = TRUE)
    rPer_low <- round(rCount_low / nrow(df), 4)
    lowPer <- c(lowPer, rPer_low)

    rCount_neu <- sum(df[[col]] == 3, na.rm = TRUE)
    rPer_neu <- round(rCount_neu / nrow(df), 4)
    neuPer <- c(neuPer, rPer_neu)

    rCount_high <- sum(df[[col]] %in% c(4, 5), na.rm = TRUE)
    rPer_high <- round(rCount_high / nrow(df), 4)
    highPer <- c(highPer, rPer_high)
  }

  df_rate <- data.frame(
    Service = service_cols,
    Low = lowPer,
    Neutral = neuPer,
    High = highPer
  )

  df_rate_long <- df_rate %>%
    pivot_longer(cols = c("Low", "Neutral", "High"), names_to = "Level", values_to = "Proportion")

  plot <- ggplot(df_rate_long, aes(x = reorder(Service, Proportion, sum), y = Proportion, fill = Level)) +
    geom_bar(stat = "identity", position = "stack") +
    coord_flip() +
    labs(x = "Services", y = "Rate Proportion", title = paste(group, "Service Rate Proportion")) +
    scale_fill_manual(values = c("Low" = "lightgrey", "Neutral" = "skyblue", "High" = "pink")) +
    geom_text(aes(label = sprintf("%.2f", Proportion * 100)), position = position_stack(vjust = 0.5), size = 3) +
    theme_minimal() +
    theme(legend.title = element_blank())

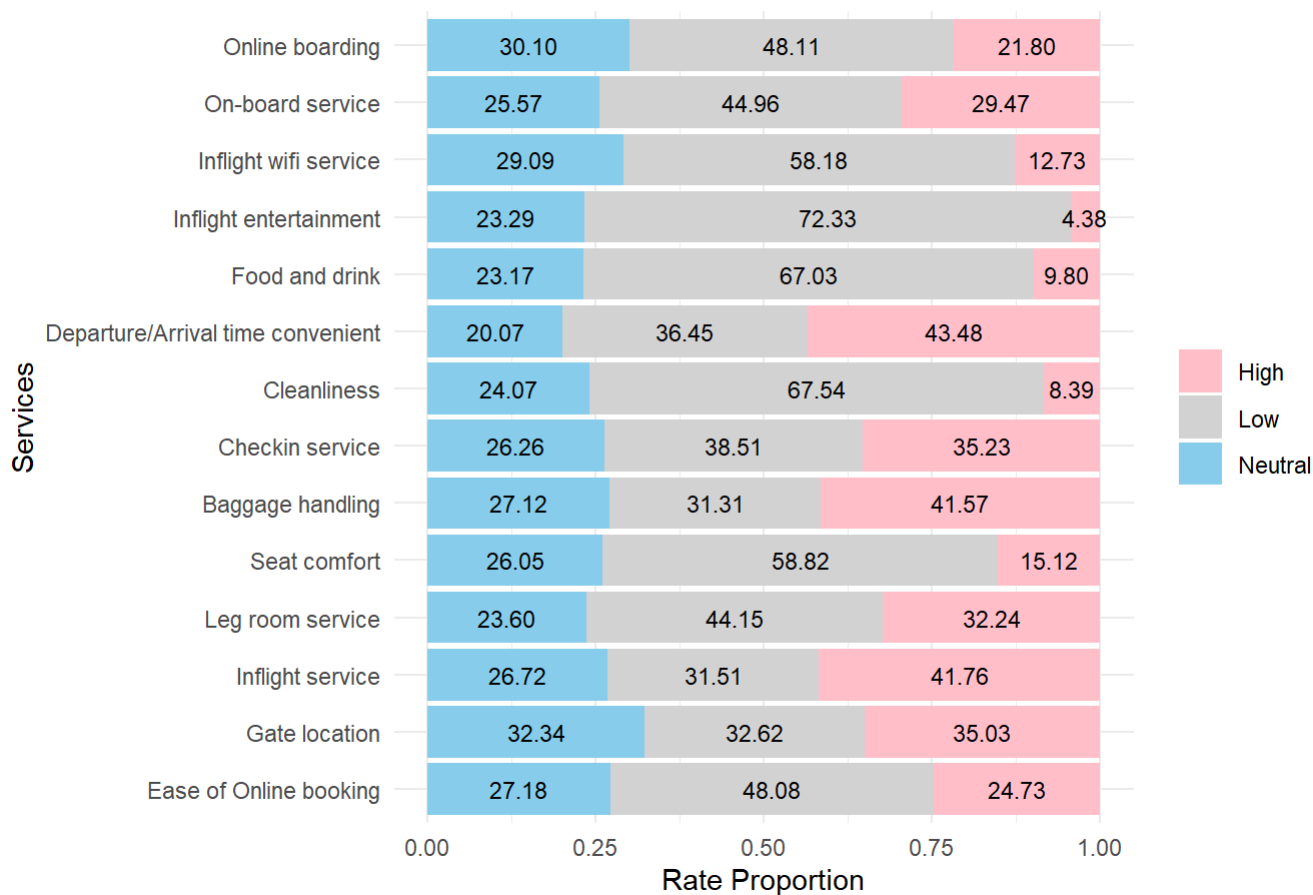
  return(plot)
}
```



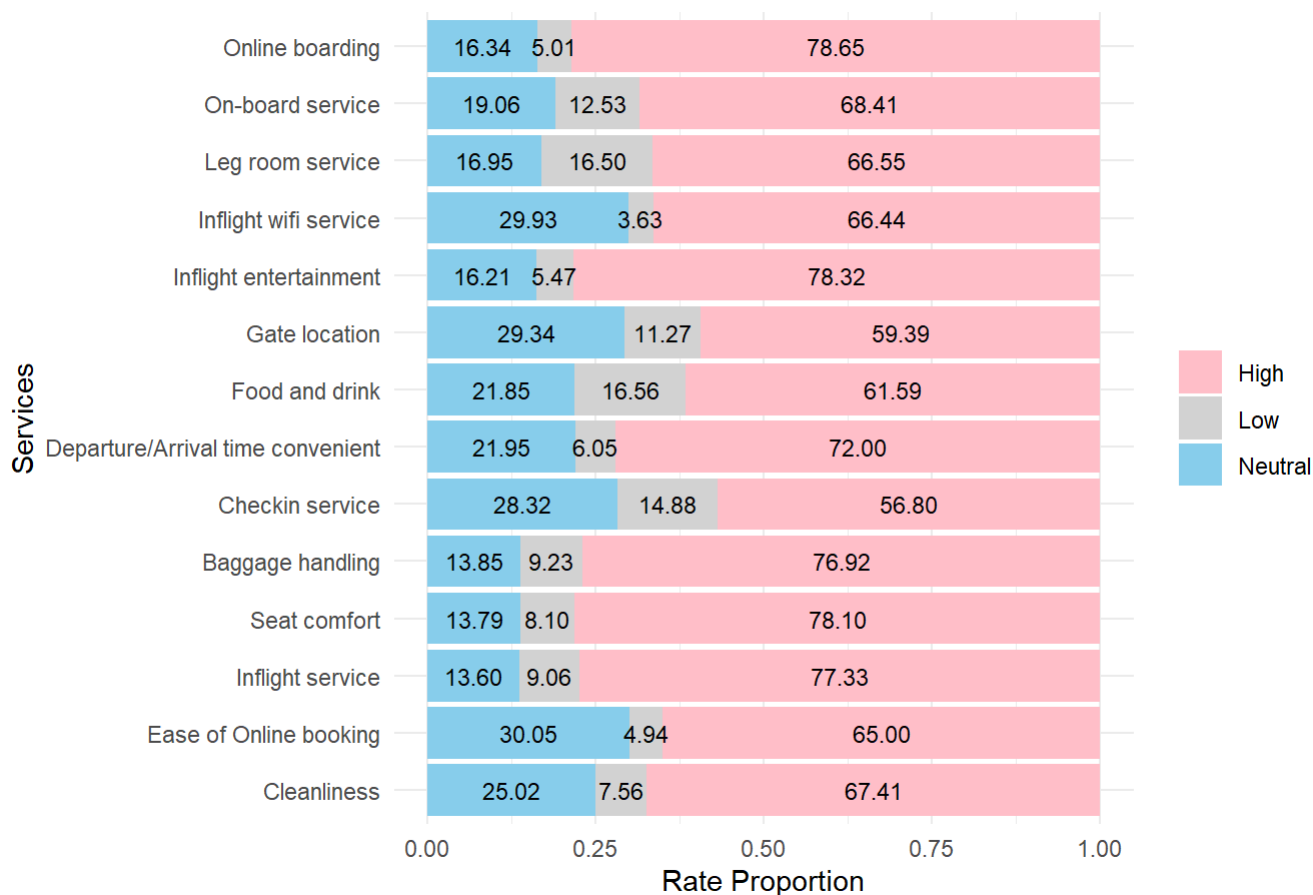
```
# 生成所有群體的圖形
df_split <- split(data_cleaned, data_cleaned$Cluster)
#上面那行是將資料分為三個子資料框，方便之後分開畫三個圖
plots <- lapply(names(df_split), function(Cluster) {
  plot_satisfaction_proportion(df_split[[Cluster]], Cluster)
})

# 顯示圖形 (逐一顯示)
lapply(plots, print)
```

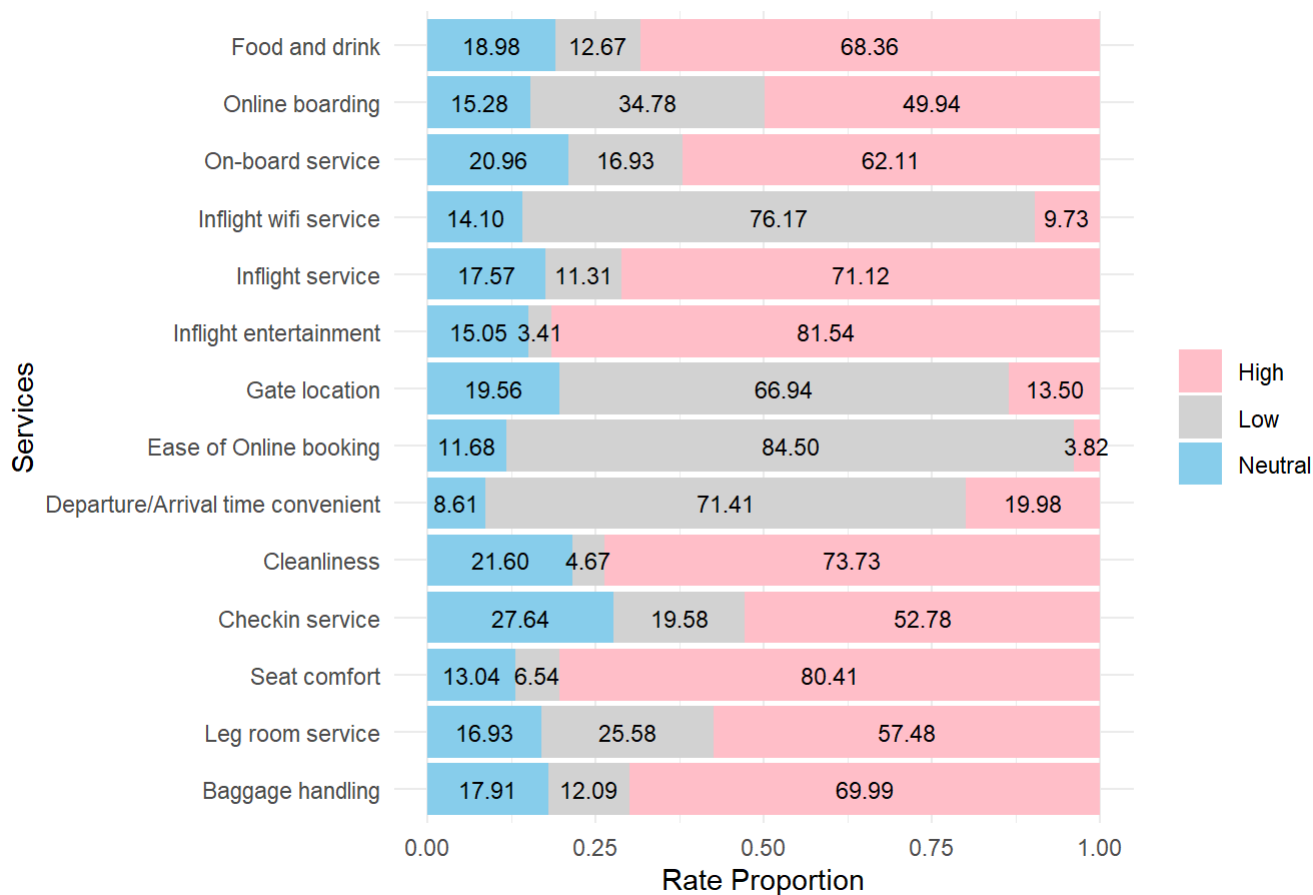
1 Service Rate Proportion



2 Service Rate Proportion

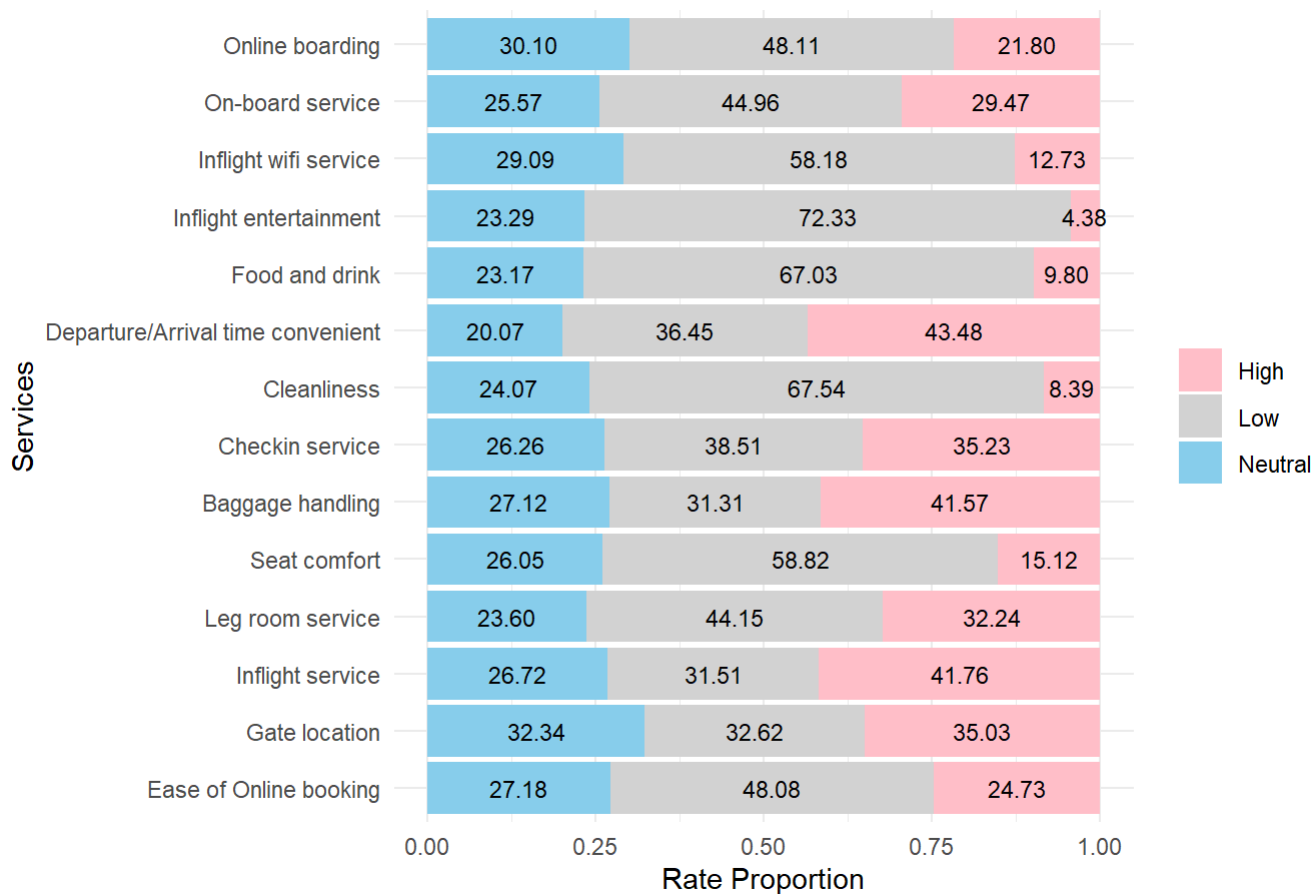


3 Service Rate Proportion



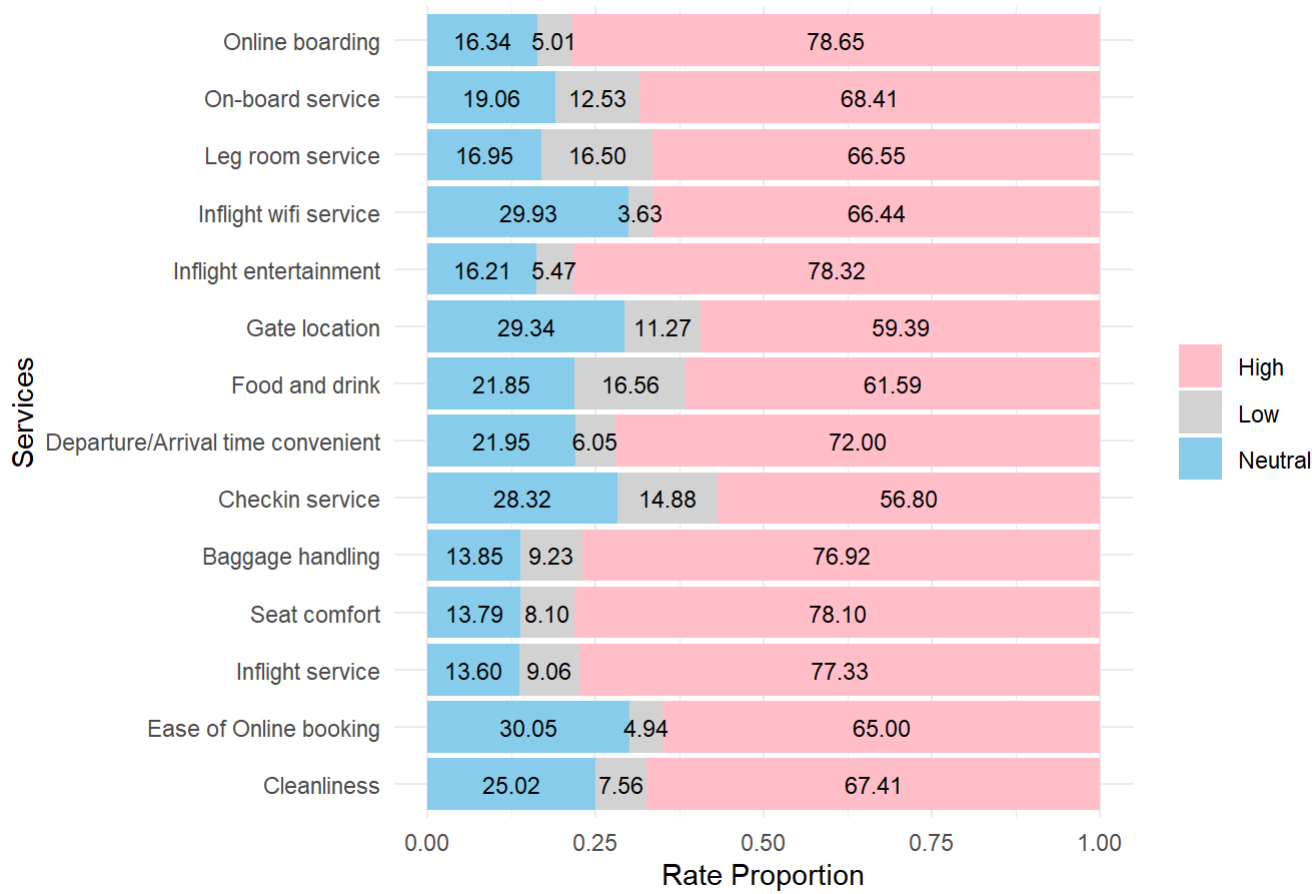
[[1]]

1 Service Rate Proportion



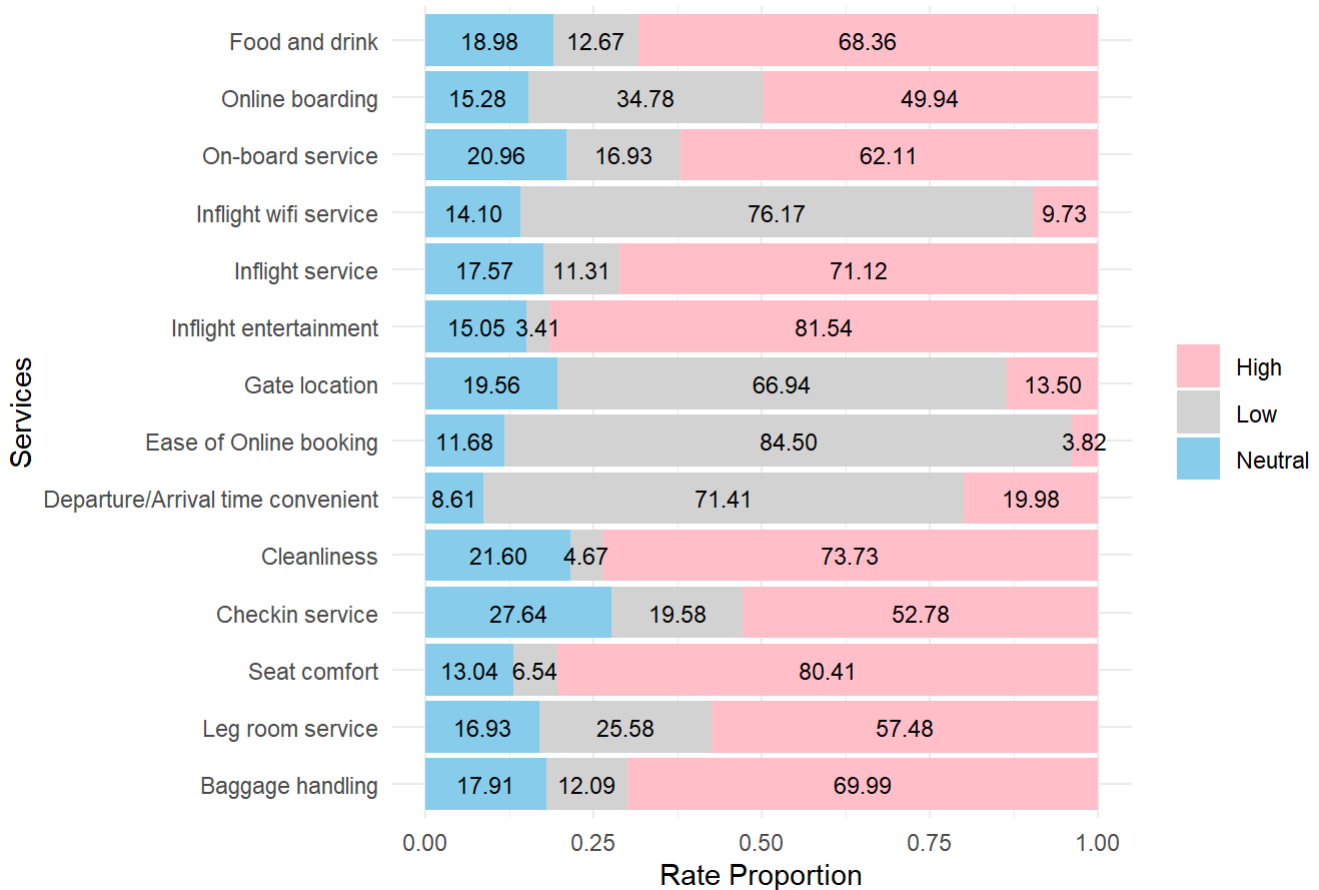
```
##  
## [[2]]
```

2 Service Rate Proportion



```
##  
## [[3]]
```

3 Service Rate Proportion



Q1.Predict passenger satisfaction

任選1種監督式學習方法配適模型，預測滿意度satisfaction (2類：滿意、中立 或 不滿意)。

找出重要變數：哪些因素影響客戶滿意度。

```
#install.packages("MASS")
library(MASS)
```

```
## Warning: 套件 'MASS' 是用 R 版本 4.4.2 來建造的
```

```
##
## 載入套件：'MASS'
```

```
## 下列物件被遮斷自 'package:dplyr':
##
## select
```

```
#install.packages("randomForest")
library(randomForest)# Q1.Predict passenger satisfaction
```

```
## Warning: 套件 'randomForest' 是用 R 版本 4.4.2 來建造的
```

```
## randomForest 4.7-1.2
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##  
## 載入套件: 'randomForest'
```

```
## 下列物件被遮斷自 'package:dplyr':  
##  
## combine
```

```
## 下列物件被遮斷自 'package:ggplot2':  
##  
## margin
```

```
# 隨機分割訓練集和測試集  
# 清理變數名稱  
colnames(data_cleaned) <- make.names(colnames(data_cleaned), unique = TRUE)  
print(colnames(data_cleaned))
```

```
## [1] "...1" "id"  
## [3] "Gender" "Customer.Type"  
## [5] "Age" "Type.of.Travel"  
## [7] "Class" "Flight.Distance"  
## [9] "Inflight.wifi.service" "Departure.Arrival.time.convenient"  
## [11] "Ease.of.Online.booking" "Gate.location"  
## [13] "Food.and.drink" "Online.boarding"  
## [15] "Seat.comfort" "Inflight.entertainment"  
## [17] "On.board.service" "Leg.room.service"  
## [19] "Baggage.handling" "Checkin.service"  
## [21] "Inflight.service" "Cleanliness"  
## [23] "Departure.Delay.in.Minutes" "Arrival.Delay.in.Minutes"  
## [25] "satisfaction" "Cluster"
```

```
trainI <- sample(1:nrow(data_cleaned), 51797)  
traind <- data_cleaned[trainI,]  
testd <- data_cleaned[-trainI,]
```

```
# 選擇需要的特徵欄位，並將滿意度變數轉為因子類型  
traind_selected <- traind[, c(9:22, 25)] # 滿意度在第25欄  
testd_selected <- testd[, c(9:22, 25)]  
  
traind_selected$satisfaction <- as.factor(traind_selected$satisfaction)  
testd_selected$satisfaction <- as.factor(testd_selected$satisfaction)
```

```
# 確認資料結構  
str(traind_selected)
```

```
## tibble [51,797 × 15] (S3: tbl_df/tbl/data.frame)
## $ Inflight.wifi.service      : num [1:51797] 3 1 4 4 1 3 3 5 2 4 ...
## $ Departure.Arrival.time.convenient: num [1:51797] 4 1 4 3 1 3 5 3 2 4 ...
## $ Ease.of.Online.booking    : num [1:51797] 3 1 4 4 1 3 3 3 2 4 ...
## $ Gate.location             : num [1:51797] 4 1 4 4 3 3 3 3 2 4 ...
## $ Food.and.drink            : num [1:51797] 3 1 2 2 3 2 5 3 2 4 ...
## $ Online.boarding           : num [1:51797] 3 4 3 2 1 5 3 5 2 4 ...
## $ Seat.comfort              : num [1:51797] 3 4 5 4 1 4 5 2 2 4 ...
## $ Inflight.entertainment    : num [1:51797] 3 2 5 5 3 4 5 5 2 4 ...
## $ On.board.service          : num [1:51797] 4 4 5 5 3 4 4 5 2 4 ...
## $ Leg.room.service          : num [1:51797] 5 3 5 5 2 4 5 5 2 3 ...
## $ Baggage.handling          : num [1:51797] 4 4 5 5 1 4 5 5 4 4 ...
## $ Checkin.service           : num [1:51797] 4 4 5 3 2 3 3 3 2 5 ...
## $ Inflight.service          : num [1:51797] 5 4 5 5 3 4 4 5 4 5 ...
## $ Cleanliness               : num [1:51797] 3 4 4 4 3 5 5 5 2 4 ...
## $ satisfaction              : Factor w/ 2 levels "neutral or dissatisfied",...: 1 1
2 2 1 2 1 2 1 2 ...
## - attr(*, "na.action")= 'omit' Named int [1:310] 214 1125 1530 2005 2109 2486 2631 3622 4
042 4491 ...
## ...- attr(*, "names")= chr [1:310] "214" "1125" "1530" "2005" ...
```

```
str(testd_selected)
```

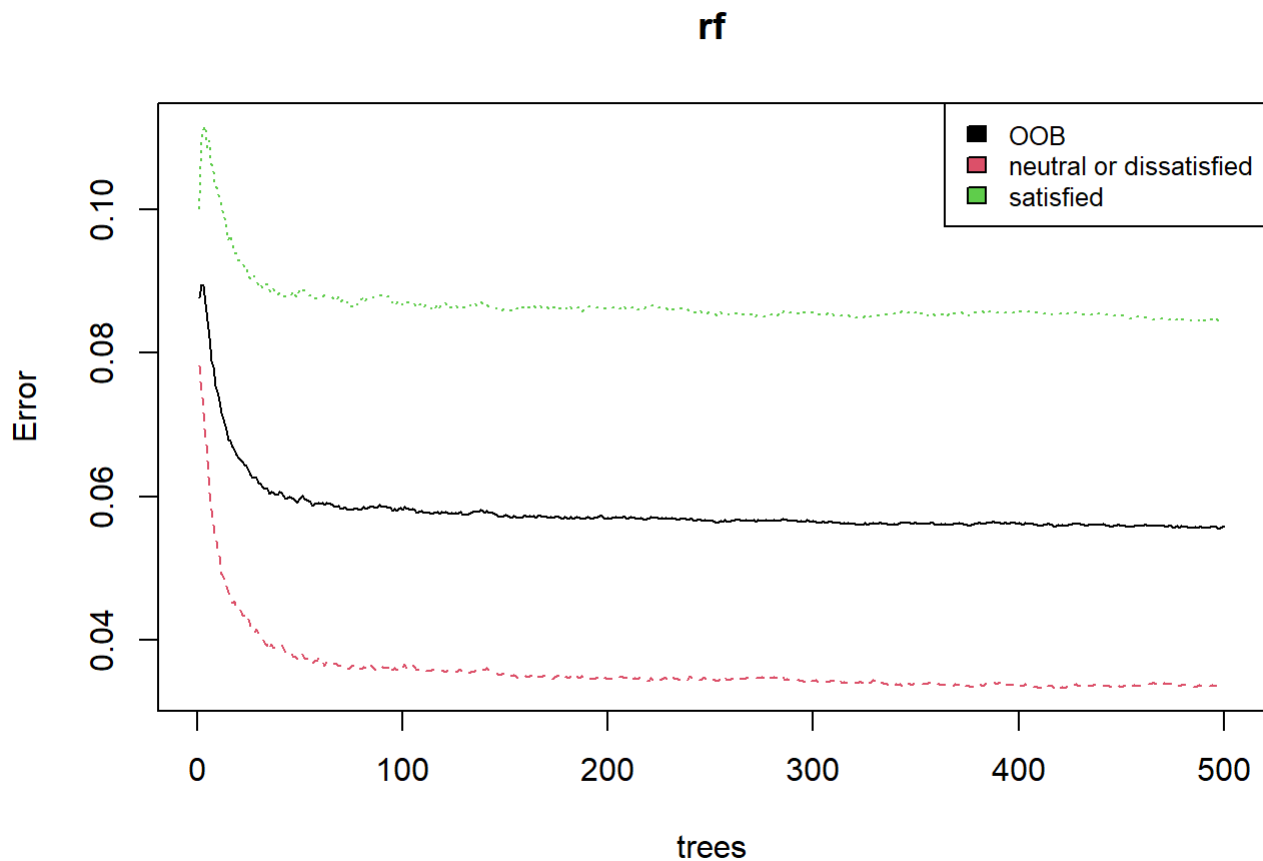
```
## tibble [51,797 × 15] (S3: tbl_df/tbl/data.frame)
## $ Inflight.wifi.service      : num [1:51797] 3 2 3 4 1 2 1 4 3 2 ...
## $ Departure.Arrival.time.convenient: num [1:51797] 2 5 4 3 2 4 4 2 2 1 ...
## $ Ease.of.Online.booking    : num [1:51797] 3 5 2 4 2 2 4 4 3 2 ...
## $ Gate.location             : num [1:51797] 3 5 1 4 2 2 4 3 2 3 ...
## $ Food.and.drink            : num [1:51797] 1 2 1 5 4 1 1 4 2 4 ...
## $ Online.boarding           : num [1:51797] 3 2 2 5 3 2 1 4 3 2 ...
## $ Seat.comfort              : num [1:51797] 1 2 1 5 3 1 1 4 2 1 ...
## $ Inflight.entertainment    : num [1:51797] 1 2 1 5 1 1 1 4 2 4 ...
## $ On.board.service          : num [1:51797] 1 2 3 5 1 1 1 4 4 2 ...
## $ Leg.room.service          : num [1:51797] 5 5 4 5 2 2 1 5 3 1 ...
## $ Baggage.handling          : num [1:51797] 3 3 4 5 1 5 3 2 2 4 ...
## $ Checkin.service           : num [1:51797] 1 1 4 4 4 5 4 2 2 1 ...
## $ Inflight.service          : num [1:51797] 4 4 4 5 1 5 4 2 1 3 ...
## $ Cleanliness               : num [1:51797] 1 2 1 4 2 1 1 4 2 4 ...
## $ satisfaction              : Factor w/ 2 levels "neutral or dissatisfied",...: 1 1
1 2 1 1 1 2 1 1 ...
## - attr(*, "na.action")= 'omit' Named int [1:310] 214 1125 1530 2005 2109 2486 2631 3622 4
042 4491 ...
## ...- attr(*, "names")= chr [1:310] "214" "1125" "1530" "2005" ...
```

```
# 建立隨機森林模型
```

```
rf <- randomForest(satisfaction ~ ., data = traind_selected, importance = TRUE)
print(rf)
```

```
##
## Call:
##  randomForest(formula = satisfaction ~ ., data = traind_selected,      importance = TRUE)
##              Type of random forest: classification
##              Number of trees: 500
## No. of variables tried at each split: 3
##
##      OOB estimate of  error rate: 5.59%
## Confusion matrix:
##              neutral or dissatisfied satisfied class.error
## neutral or dissatisfied      28385      990  0.03370213
## satisfied                  1903    20519  0.08487200
```

```
# 視覺化錯誤率隨著樹數的變化
plot(rf)
legend("topright", colnames(rf$err.rate), col = 1:4, cex = 0.8, fill = 1:4)
```



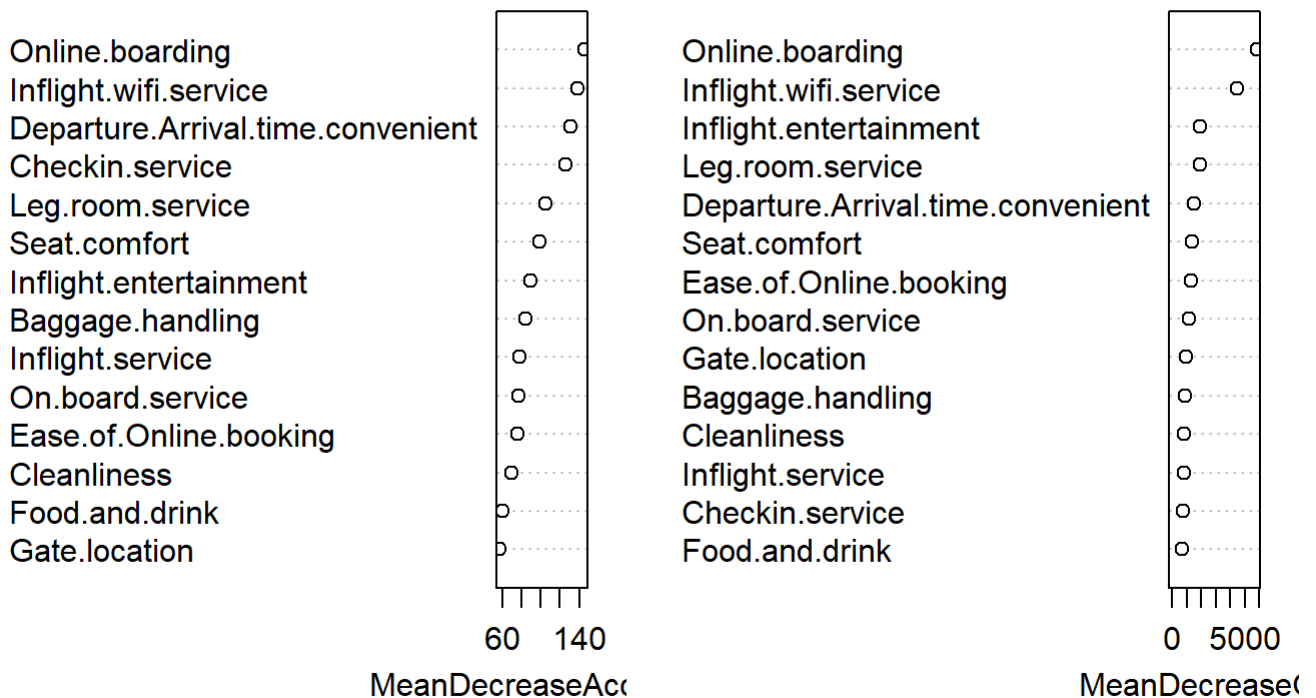
```
# 評估變數重要性
importance(rf)
```



```
##                                neutral or dissatisfied satisfied
## Inflight.wifi.service          146.33129  98.05745
## Departure.Arrival.time.convenient 107.59545 108.48301
## Ease.of.Online.booking         66.02662  53.47760
## Gate.location                  24.60070  55.07099
## Food.and.drink                 40.53854  51.26370
## Online.boarding                114.86419 140.67822
## Seat.comfort                   88.93986  65.84248
## Inflight.entertainment         61.91215  77.26274
## On.board.service               61.44378  60.55205
## Leg.room.service              77.75581  87.71877
## Baggage.handling              84.82988  55.60058
## Checkin.service               114.33969  52.50394
## Inflight.service              74.96390  52.44417
## Cleanliness                   54.96988  51.41146
##                                MeanDecreaseAccuracy MeanDecreaseGini
## Inflight.wifi.service          138.20685      4478.6967
## Departure.Arrival.time.convenient 130.91415      1486.4767
## Ease.of.Online.booking         75.57719      1291.5536
## Gate.location                  57.14426       969.5097
## Food.and.drink                 59.83003       694.2901
## Online.boarding                145.45804      5827.9352
## Seat.comfort                   98.54041      1360.3793
## Inflight.entertainment         89.46290      1921.5451
## On.board.service               76.53113      1134.0046
## Leg.room.service              104.80274      1880.1505
## Baggage.handling              83.89769       864.7709
## Checkin.service               125.61768       722.3460
## Inflight.service              77.44926       783.1376
## Cleanliness                   69.53207       828.1477
```

```
varImpPlot(rf)
```

rf



*#Mean Decrease Accuracy - How much the model accuracy decreases if we drop that variable.
 #Mean Decrease Gini - Measure of variable importance based on the Gini impurity index used for the calculation of splits in trees.*

```
# 預測測試集的滿意度
pred <- predict(rf, newdata = testd_selected)
```

```
# 混淆矩陣：實際值與預測值的對比
conf_matrix <- table(Real = testd_selected$satisfaction, Predict = pred)
```

```
#計算分數
#準確率 (Accuracy)
accuracy <- sum(diag(conf_matrix)) / sum(conf_matrix) # diag(conf_matrix)是左上那一格
print(paste("Accuracy:", round(accuracy, 4)))
```

```
## [1] "Accuracy: 0.9451"
```

```
#精確率 (Precision) 和 召回率 (Recall)
precision <- diag(conf_matrix) / colSums(conf_matrix)
recall <- diag(conf_matrix) / rowSums(conf_matrix)
print(data.frame(Class = rownames(conf_matrix), Precision = precision, Recall = recall))
```

```
##                                Class Precision    Recall
## neutral or dissatisfied neutral or dissatisfied 0.9389298 0.9658277
## satisfied                                satisfied 0.9536862 0.9180423
```

```
#F1 分數
```

```
f1_score <- 2 * (precision * recall) / (precision + recall)
print(data.frame(Class = rownames(conf_matrix), F1_Score = f1_score))
```

```
##                                Class  F1_Score
## neutral or dissatisfied neutral or dissatisfied 0.9521888
## satisfied                                satisfied 0.9355248
```

```
#混淆矩陣可視化
#install.packages("caret")
library(caret)
```

```
## Warning: 套件 'caret' 是用 R 版本 4.4.2 來建造的
```

```
## 載入需要的套件 : lattice
```

```
##
## 載入套件 : 'caret'
```

```
## 下列物件被遮斷自 'package:purrr':
##
##      lift
```

```
confusionMatrix(pred, testd_selected$satisfaction)
```

```
## Confusion Matrix and Statistics
##
##               Reference
## Prediction      neutral or dissatisfied satisfied
## neutral or dissatisfied      28320      1842
## satisfied                  1002      20633
##
##               Accuracy : 0.9451
##               95% CI : (0.9431, 0.947)
##               No Information Rate : 0.5661
##               P-Value [Acc > NIR] : < 2.2e-16
##
##               Kappa : 0.8877
##
## Mcnemar's Test P-Value : < 2.2e-16
##
##               Sensitivity : 0.9658
##               Specificity : 0.9180
##               Pos Pred Value : 0.9389
##               Neg Pred Value : 0.9537
##               Prevalence : 0.5661
##               Detection Rate : 0.5467
##               Detection Prevalence : 0.5823
##               Balanced Accuracy : 0.9419
##
##               'Positive' Class : neutral or dissatisfied
##
```

```
library(reshape2)
```

```
## Warning: 套件 'reshape2' 是用 R 版本 4.4.2 來建造的
```

```
##
## 載入套件：'reshape2'
```

```
## 下列物件被遮斷自 'package:tidyr':
##
##      smiths
```

```
library(ggplot2)
conf_matrix_melt <- melt(conf_matrix)
ggplot(data = conf_matrix_melt, aes(x = Real, y = Predict, fill = value)) +
  geom_tile() +
  geom_text(aes(label = value), color = "white") +
  scale_fill_gradient(low = "blue", high = "red") +
  labs(title = "Confusion Matrix", x = "Actual", y = "Predicted") +
  theme_minimal()
```

Confusion Matrix

