

HW4

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讀取資料：知名度與支持度

```
library(dplyr)
library(ggplot2)
library(haven)

data <- read_sav("poll.sav")
data_north<-data[which(data[,1]==1),]
data_west<-data[which(data[,1]==2),]#
data_north<-data_north[,-3]
data_west<-data_west[,-2]

north_multiple_answer<-data_north[,c(3:10)]
west_multiple_answer<-data_west[,c(3:10)]
multiple_answer<-data[,c(4:11)]
sum(north_multiple_answer == 3)
```

[1] 319

```

#
#
candidate_ids <- 1:10
regions <- c(" ", " ") #
age<-c(1:5)
edu<-c(1:5)
sex<-c(1:2)
#
popularity <- data.frame(matrix(nrow = length(candidate_ids),
                                ncol = length(regions)+
                                    length(age)+
                                    length(edu)+
                                    length(sex)))
rownames(popularity) <- paste("candidate", candidate_ids, sep = "")
colnames(popularity) <- c(regions,
                           paste("age", age, sep = ""),
                           paste("education", edu, sep = ""),
                           paste("sex", sex, sep = ""))

popularity[, " "] <- sapply(candidate_ids,
function(i) sum(north_multiple_answer == i))

popularity[, " "] <- sapply(candidate_ids,
function(i) sum(west_multiple_answer == i))

for (v in 1:5) {
  col_name <- paste("age", v, sep = "")
  popularity[, col_name] <- sapply(candidate_ids,
function(i) sum((multiple_answer == i) & (data$v6 == v)))
}

for (e in 1:5) {
  col_name <- paste("education", e, sep = "")
  popularity[, col_name] <- sapply(candidate_ids,
function(i) sum((multiple_answer == i) & (data$v7 == e)))
}

for (s in 1:2) {
  col_name <- paste("sex", s, sep = "")
  popularity[, col_name] <- sapply(candidate_ids,
function(i) sum((multiple_answer == i) & (data$v8 == s)))
}

```

```

#
library(reshape2)
popularity$candidate<-paste("candidate",1:10,sep="")

popularity <- popularity%>%
  mutate(candidate=case_when(
    candidate == "candidate1" ~ "1 ",
    candidate == "candidate2" ~ "2 ",
    candidate == "candidate3" ~ "3 ",
    candidate == "candidate4" ~ "4 ",
    candidate == "candidate5" ~ "5 ",
    candidate == "candidate6" ~ "6 ",
    candidate == "candidate7" ~ "7 ",
    candidate == "candidate8" ~ "8 ",
    candidate == "candidate9" ~ "9 ",
    candidate == "candidate10" ~ "10 ",
    TRUE ~ as.character(candidate)
  ))

popularity_long_region <- melt(popularity,
                               id.vars = "candidate",
                               measure.vars = c(" ", " "),
                               variable.name = "region",
                               value.name = "popularity")

# age
popularity_long_age <- melt(popularity, id.vars = "candidate",
                           measure.vars = paste("age", 1:5, sep = ""),
                           variable.name = "age",
                           value.name = "popularity")
popularity_long_age <-popularity_long_age%>%
  mutate(age = case_when(
    age == "age1" ~ "20-29 ",
    age == "age2" ~ "30-39 ",
    age == "age3" ~ "40-49 ",
    age == "age4" ~ "50-59 ",
    age == "age5" ~ "60 ",
    TRUE ~ as.character(age)
  ))

# education
popularity_long_education <- melt(popularity,
                                  id.vars = "candidate",

```

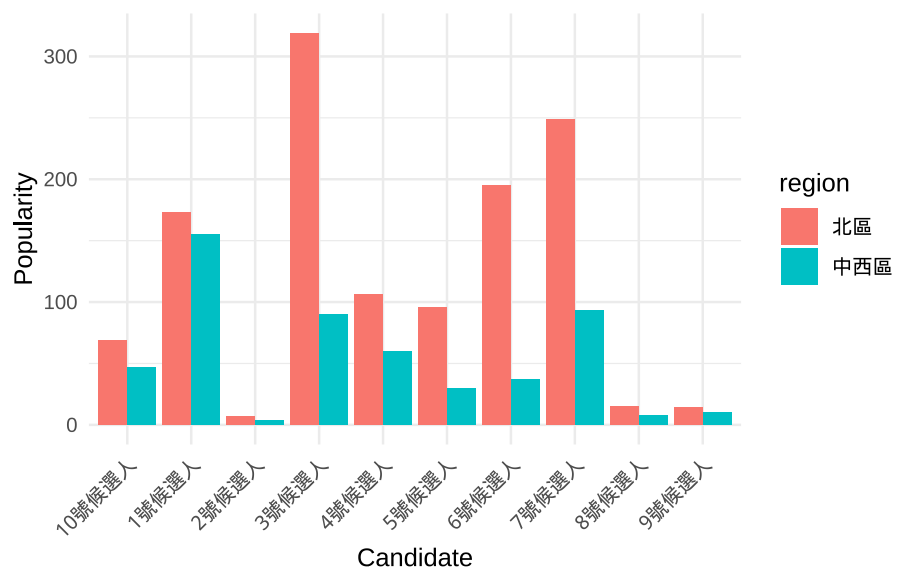
```

        measure.vars = paste("education", 1:5, sep = ""),
        variable.name = "education",
        value.name = "popularity")
popularity_long_education <-popularity_long_education%>%
  mutate(education = case_when(
    education == "education1" ~ "1:  ",
    education == "education2" ~ "2:  ",
    education == "education3" ~ "3:  ",
    education == "education4" ~ "4:  ",
    education == "education5" ~ "5:  ",
    TRUE ~ as.character(education)
  ))

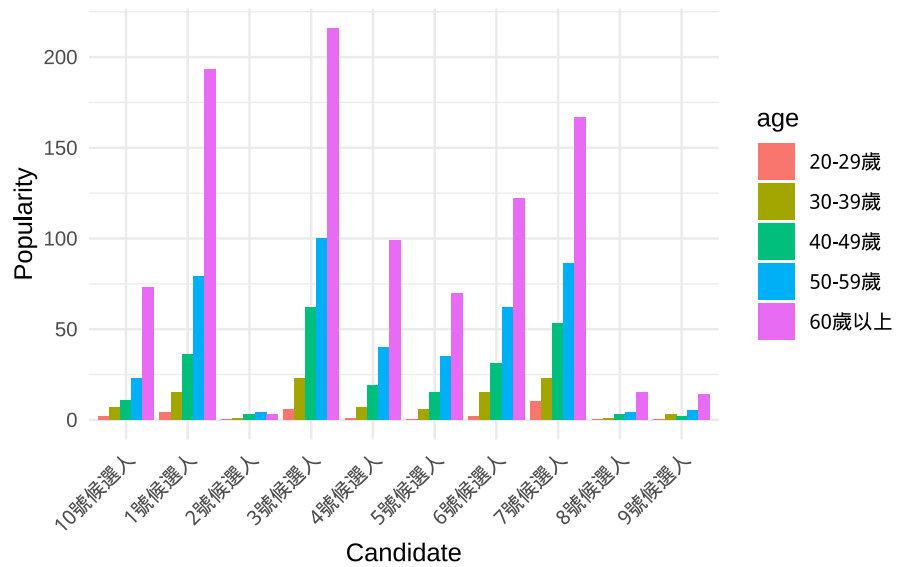
# sex
popularity_long_sex <- melt(popularity,
  id.vars = "candidate",
  measure.vars = paste("sex", 1:2, sep = ""),
  variable.name = "sex",
  value.name = "popularity")
popularity_long_sex <-popularity_long_sex%>%
  mutate(sex = case_when(
    sex == "sex1" ~ "1:  ",
    sex == "sex2" ~ "2:  ",
    TRUE ~ as.character(sex)
  ))

#
ggplot(popularity_long_region,
  aes(x = candidate, y = popularity, fill = region)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Popularity") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

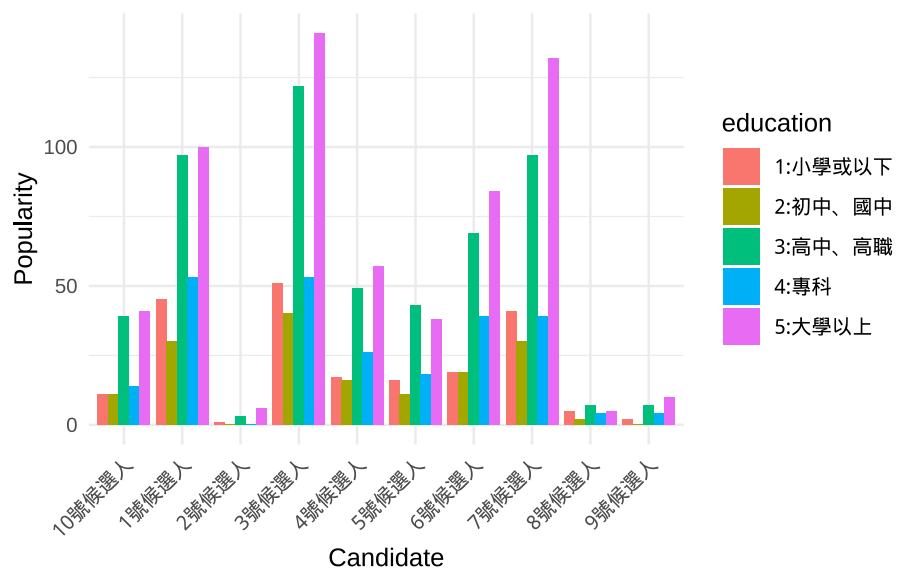
```



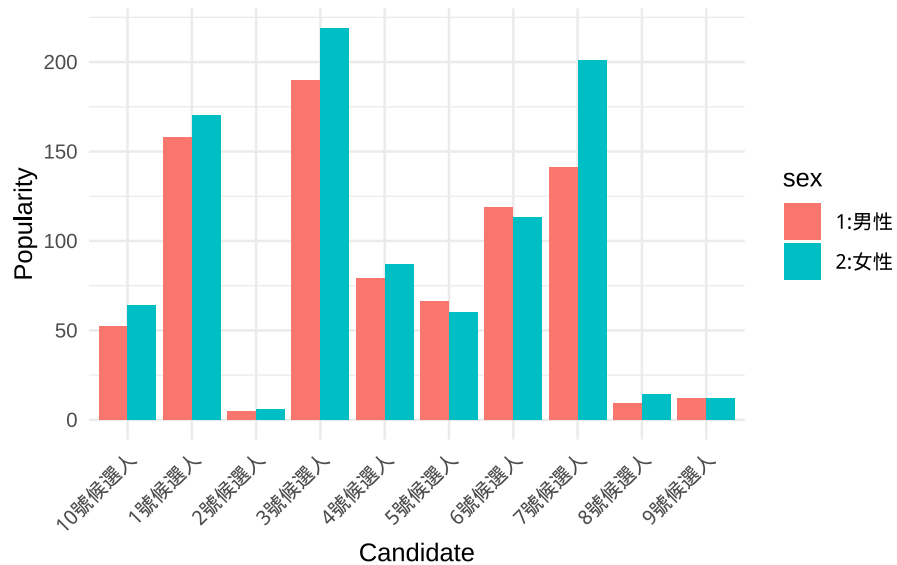
```
ggplot(popularity_long_age,
       aes(x = candidate, y = popularity, fill = age)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Popularity") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
ggplot(popularity_long_education,
       aes(x = candidate, y = popularity, fill = education)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Popularity") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
ggplot(popularity_long_sex,
       aes(x = candidate, y = popularity, fill = sex)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Popularity") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



#Support rating

```

candidate_ids <- 1:10
regions <- c(" ", " ")
age<-c(1:6)
edu<-c(1:5)
sex<-c(1:2)
#
Support_rating <- data.frame(matrix(nrow = length(candidate_ids),
                                   ncol = length(regions)+
                                   length(age)+
                                   length(edu)+
                                   length(sex)))
rownames(Support_rating) <- paste("candidate", candidate_ids, sep = "")
colnames(Support_rating) <- c(regions,paste("age", age, sep = ""),
                             paste("education", edu, sep = ""),
                             paste("sex", sex, sep = ""))

Support_rating[, " "] <- sapply(candidate_ids ,
function(i) sum(data_north$v5 == i))

Support_rating[, " "] <- sapply(candidate_ids,
function(i) sum(data_west$v5 == i))

```



```

for (v in 1:5) {
  col_name <- paste("age", v, sep = "")
  Support_rating[, col_name] <- sapply(candidate_ids,
function(i) sum((data$v5 == i) & (data$v6 == v)))
}

for (e in 1:5) {
  col_name <- paste("education", e, sep = "")
  Support_rating[, col_name] <- sapply(candidate_ids,
function(i) sum((data$v5 == i) & (data$v7 == e)))
}

for (s in 1:2) {
  col_name <- paste("sex", s, sep = "")
  Support_rating[, col_name] <- sapply(candidate_ids,
function(i) sum((data$v5 == i) & (data$v8 == s)))
}

Support_rating$candidate<-paste("candidate",1:10,sep="")
Support_rating <- Support_rating%>%
mutate(candidate=case_when(
  candidate == "candidate1" ~ "1 ",
  candidate == "candidate2" ~ "2 ",
  candidate == "candidate3" ~ "3 ",
  candidate == "candidate4" ~ "4 ",
  candidate == "candidate5" ~ "5 ",
  candidate == "candidate6" ~ "6 ",
  candidate == "candidate7" ~ "7 ",
  candidate == "candidate8" ~ "8 ",
  candidate == "candidate9" ~ "9 ",
  candidate == "candidate10" ~ "10 ",
  TRUE ~ as.character(candidate)
))

#
Support_rating_long_region <- melt(Support_rating, id.vars = "candidate",
  measure.vars = c(" ", " "),
  variable.name = "region",
  value.name = "Support_rating")

Support_rating_long_age <- melt(Support_rating, id.vars = "candidate",
  measure.vars = paste("age", 1:5, sep = ""),
  variable.name = "age",

```

```

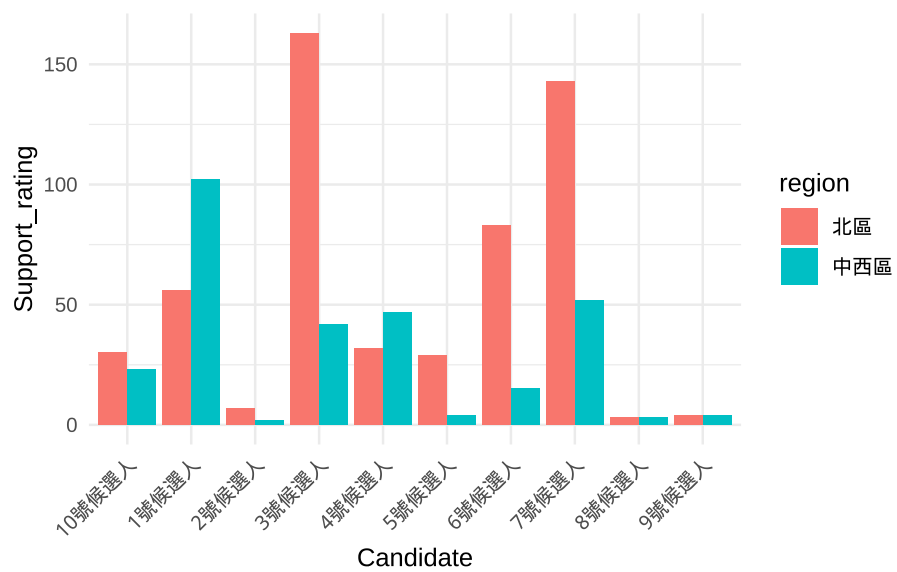
                                value.name = "Support_rating")
Support_rating_long_age<- Support_rating_long_age%>%
  mutate(age = case_when(
    age == "age1" ~ "20-29 ",
    age == "age2" ~ "30-39 ",
    age == "age3" ~ "40-49 ",
    age == "age4" ~ "50-59 ",
    age == "age5" ~ "60  ",
    TRUE ~ as.character(age)
  ))

Support_rating_long_education <- melt(Support_rating, id.vars = "candidate",
                                     measure.vars = paste("education", 1:5, sep = ""),
                                     variable.name = "education",
                                     value.name = "Support_rating")
Support_rating_long_education <-Support_rating_long_education%>%
  mutate(education = case_when(
    education == "education1" ~ "1:  ",
    education == "education2" ~ "2:  ",
    education == "education3" ~ "3:  ",
    education == "education4" ~ "4:  ",
    education == "education5" ~ "5:  ",
    TRUE ~ as.character(education)
  ))

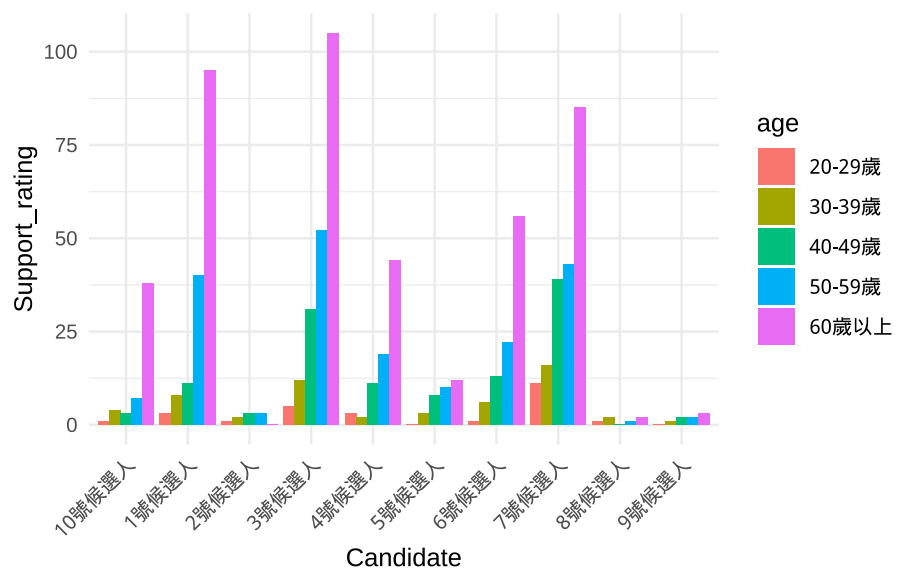
Support_rating_long_sex <- melt(Support_rating, id.vars = "candidate",
                              measure.vars = paste("sex", 1:2, sep = ""),
                              variable.name = "sex",
                              value.name = "Support_rating")
Support_rating_long_sex <- Support_rating_long_sex%>%
  mutate(sex = case_when(
    sex == "sex1" ~ "1:  ",
    sex == "sex2" ~ "2:  ",
    TRUE ~ as.character(sex)
  ))

ggplot(Support_rating_long_region,
       aes(x = candidate, y = Support_rating, fill = region)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Support_rating") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

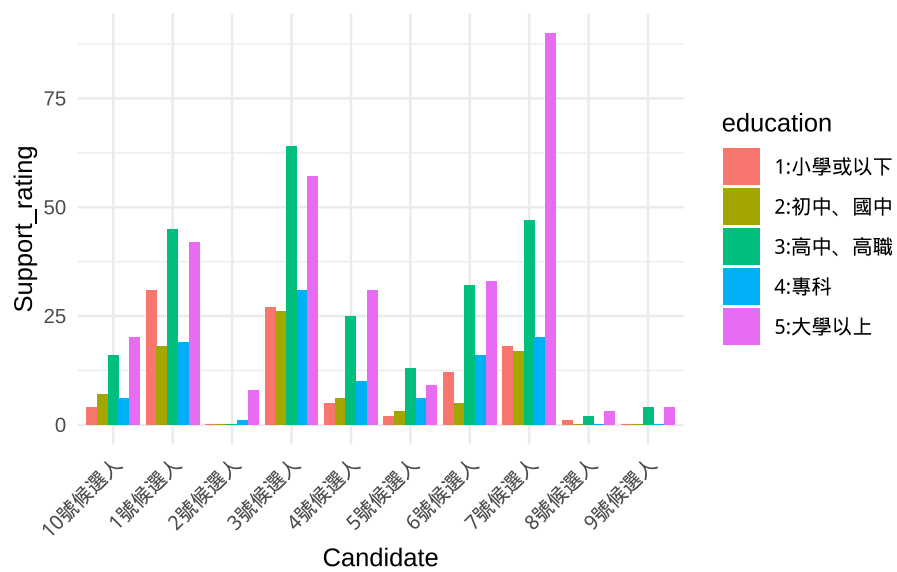
```



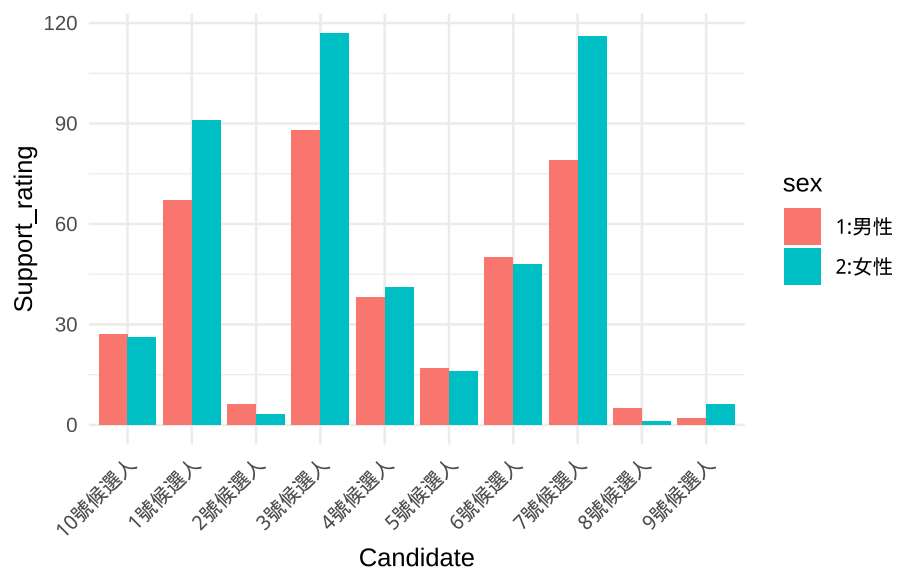
```
ggplot(Support_rating_long_age,
  aes(x = candidate, y = Support_rating, fill = age)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Support_rating") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
ggplot(Support_rating_long_education,
  aes(x = candidate, y = Support_rating, fill = education)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Support_rating") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

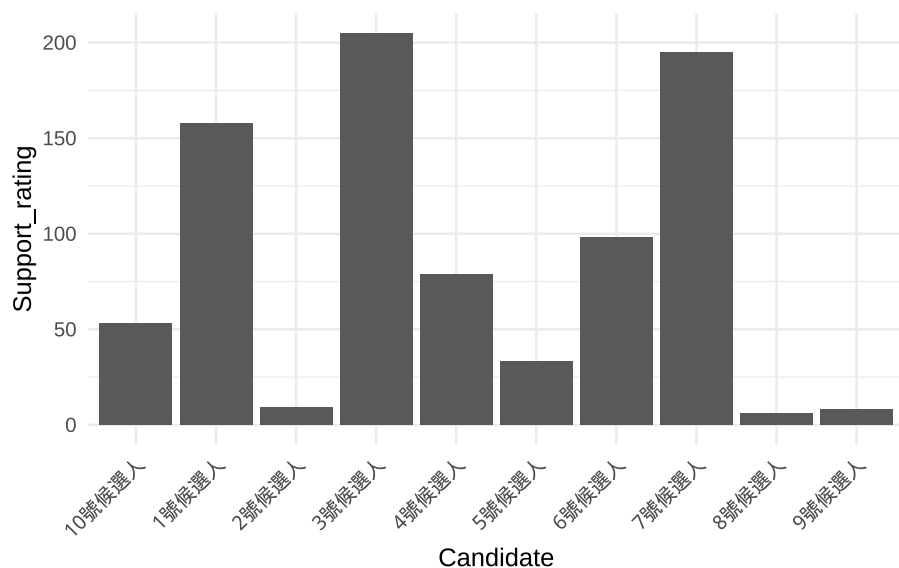


```
ggplot(Support_rating_long_sex,
  aes(x = candidate, y = Support_rating, fill = sex)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Candidate", y = "Support_rating") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
total_supporting<-data.frame(matrix(nrow = length(candidate_ids), ncol = 1))
colnames(total_supporting)<-"Support_rating"
total_supporting$candidate<-c("1 ",
                                "2 ",
                                "3 ",
                                "4 ",
                                "5 ",
                                "6 ",
                                "7 ",
                                "8 ",
                                "9 ",
                                "10 ")
total_supporting[, "Support_rating"] <- sapply(candidate_ids ,

ggplot(total_supporting, aes(x = candidate, y =Support_rating)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(, x = "Candidate", y = "Support_rating") +
  theme_minimal()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



1、知名度與支持度分析

1. 整體來看，不論是知名度或支持度，2,8,9號候選人都很沒有競爭力。
依照地區的知名度來看，北區最高是3號候選人，中西區最高是1號。
而在中西區，3,4,7,10號候選人的知名度相差不大。
2. 高齡大多知曉3號候選人，而年輕族群則較為熟識7號候選人。
3. 教育水準較高的選民較多比例會選擇3號候選人，反之較低者會傾向選7號候選人。
4. 支持度最高為3號候選人，其次為7號，但差異不大。

2、三號候選人選舉策略

觀察

知名度

1. 3號候選人在中西區的知名度低於1,7號候選人
2. 在年輕選民中（20-40間），3號候選人的知名度低於7號候選人，尤其是在20-30歲的區段。
3. 3號候選人在各個教育程度的群眾中皆較為知名，但與7號候選人的差距在高教育程度的選民時減小。

支持度

4. 教育水準較低的選民較傾向3號候選人，反之較高者會傾向選7號候選人。

5. 整體而言，3號候選人的支持度略微高於7。

結論

認為3號候選人若想在市議員選舉中增加當選的勝算，策略上應該以7號候選人作為主要競爭對手。除了穩住自己的基本盤（高齡、低教育程度）以外，應該也向年輕選民和高教育程度者多多表現自己，提升支持度。且因為調查限制為20歲，忽略了選舉時剛獲得選舉權的首投族，以此調查的趨勢來看，7號候選人可能有潛藏的票倉，需要注意。

3、第3號候選人支持率的預測模式

```
# logistic
dt<-data[,c("v1","v5","v6","v7","v8")]
colnames(dt)<-c("region","support","age","education","sex")
dt$region<-as.factor(dt$region)
dt[, "support"]<-dt[, "support"]==3
dt<-dt[-which(dt[, "age"]==6),]#
dt<-dt[-which(dt[, "education"]==95),]#
dt$age <- factor(dt$age)
dt$education<-factor(dt$education)
dt$sex<-as.factor(dt$sex)
model1<-glm(dt$support~dt$region+dt$age+dt$education+dt$sex,
            data=dt,
            family = binomial(link = "logit"))
summary(model1)
```

Call:

```
glm(formula = dt$support ~ dt$region + dt$age + dt$education +
     dt$sex, family = binomial(link = "logit"), data = dt)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-2.08096	0.55148	-3.773	0.000161	***
dt\$region2	-0.72945	0.18312	-3.983	6.79e-05	***
dt\$age2	0.30874	0.56826	0.543	0.586914	
dt\$age3	0.45620	0.51624	0.884	0.376852	
dt\$age4	0.37840	0.50210	0.754	0.451064	
dt\$age5	0.05231	0.49757	0.105	0.916275	
dt\$education2	0.54130	0.30091	1.799	0.072035	.
dt\$education3	0.37288	0.25697	1.451	0.146771	
dt\$education4	0.39770	0.30565	1.301	0.193206	
dt\$education5	-0.06315	0.28283	-0.223	0.823305	
dt\$sex2	-0.06044	0.15613	-0.387	0.698664	

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

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1225.3 on 1600 degrees of freedom
Residual deviance: 1192.7 on 1590 degrees of freedom
AIC: 1214.7

Number of Fisher Scoring iterations: 5

```
#Step method  
modell1.step <- step(modell1,direction = c("both"))
```

Start: AIC=1214.75

dt\$support ~ dt\$region + dt\$age + dt\$education + dt\$sex

	Df	Deviance	AIC
- dt\$age	4	1197.1	1211.1
- dt\$sex	1	1192.9	1212.9
<none>		1192.8	1214.8
- dt\$education	4	1201.7	1215.7
- dt\$region	1	1210.3	1230.3

Step: AIC=1211.08

dt\$support ~ dt\$region + dt\$education + dt\$sex

	Df	Deviance	AIC
- dt\$sex	1	1197.2	1209.2
<none>		1197.1	1211.1
- dt\$education	4	1206.5	1212.5
+ dt\$age	4	1192.8	1214.8
- dt\$region	1	1215.7	1227.7

Step: AIC=1209.15

dt\$support ~ dt\$region + dt\$education

	Df	Deviance	AIC
<none>		1197.2	1209.2
- dt\$education	4	1206.8	1210.8
+ dt\$sex	1	1197.1	1211.1
+ dt\$age	4	1192.9	1212.9
- dt\$region	1	1215.8	1225.8

```
summary(model1.step)
```

Call:

```
glm(formula = dt$support ~ dt$region + dt$education, family = binomial(link = "logit"),
     data = dt)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-2.0600	0.2073	-9.937	< 2e-16 ***
dt\$region2	-0.7481	0.1826	-4.096	4.2e-05 ***
dt\$education2	0.6172	0.2959	2.086	0.0370 *
dt\$education3	0.5075	0.2445	2.075	0.0379 *
dt\$education4	0.6118	0.2829	2.163	0.0306 *
dt\$education5	0.1581	0.2472	0.640	0.5224

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

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1225.3 on 1600 degrees of freedom
Residual deviance: 1197.2 on 1595 degrees of freedom
AIC: 1209.2

Number of Fisher Scoring iterations: 5

模型觀察

Logistic : Step Method

透過Step method，我們知道模型中education2,3,4與region2為顯著。
因模型預設以每個變數的1為基底，我們得到以下結論：

- region2(中西區)的係數約為-0.7481。
代表選民來自中西區時，支持三號候選人的可能性約為北區選民的
 $e^{-0.7481} = 0.473$ 倍
- education2(初中、國中)的係數約為0.6172。
代表選民最高學歷為初中或國中時，支持三號候選人的可能性約為最高學歷為小學或以下的
 $e^{0.6172} = 1.855$ 倍
- education3(高中、高職)的係數約為0.5075。
代表選民最高學歷為高中或高職時，支持三號候選人的可能性約為最高學歷為小學或以下的
 $e^{0.5075} = 1.661$ 倍
- education4(專科)的係數約為0.6118。
代表選民最高學歷為專科時，支持三號候選人的可能性約為最高學歷為小學或以下的

$$e^{0.6118} = 1.844 \text{ 倍}$$