
Analyzing Campaign Finance Data in California using SAS & R

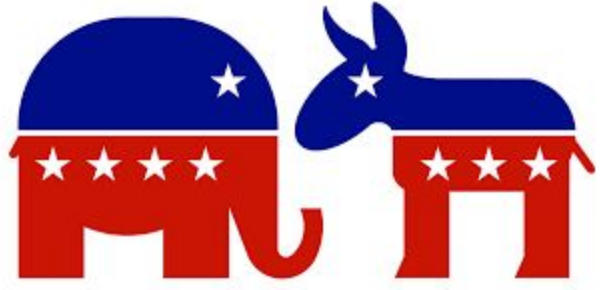
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Background Info



- In the US in every even year, Americans cast their vote in all 50 states in national, state, and local elections
- This data was collected on kaggle from the Federal Election Commission (FEC) for campaign finance data from 2019-2020 leading up to the 2020 election for all major US house of representatives candidates
- I chose to further analysis campaign finance data for my home state of California to better understand where each party was collecting their political campaign contributions from

Voting!



- Live in California
- Difference in campaign finance contributions by political party?
- Pertinent as the 2022 midterm elections occurred this month

The Data



FEC.gov

The LOGISTIC Procedure

Model Information	
Data Set	WORK.CALI
Response Variable	CAND_PTY_AFFILIATION
Number of Response Levels	3
Model	generalized logit
Optimization Technique	Newton-Raphson

Number of Observations Read	330
Number of Observations Used	330

Response Profile		
Ordered Value	CAND_PTY_AFFILIATION	Total Frequency
1	DEM	180
2	IND	5
3	REP	145

Variables ~

- **CAND_PTY_AFFILIATION** = political party
- **CAND_LOANS** = loans from the candidate
- **Receipts** = total \$ receipts for the candidate
- **TTL_INDIV_CONTRIB** = contributions made from the candidate themselves

Data Cleaning



```
proc import out=cand
  datafile="/home/u60876478/410Project/All candidates.csv"
  dbms=csv replace;
run;

proc sql;
  create table cali as
  select *
  from cand
  where CAND_OFFICE_ST = 'CA';

proc sql;
  delete
  from cali
  where CAND_PTY_AFFILIATION = 'UNK' or CAND_PTY_AFFILIATION = 'OTH'
  or CAND_PTY_AFFILIATION = 'NON' or CAND_PTY_AFFILIATION = 'GRE' or
  CAND_PTY_AFFILIATION = 'NPA' or CAND_PTY_AFFILIATION = 'NNE' or
  CAND_PTY_AFFILIATION = 'LIB';
```



```
library(dplyr)
library(tidyverse)
```

```
cali = All_candidates %>%
  rename(state = CAND_OFFICE_ST)
cali = cali %>%
  filter(state == "CA") %>%
  rename(party = CAND_PTY_AFFILIATION) %>%
  rename(receipts = TTL_RECEIPTS)
```

```
cali = cali %>%
  filter(party != "GRE")
cali = cali %>%
  filter(party != "NPA")
cali = cali %>%
  filter(party != "NON")
cali = cali %>%
  filter(party != "LIB")
cali = cali %>%
  filter(party != "OTH")
cali = cali %>%
  filter(party != "UNK")
cali = cali %>%
  filter(party != "NNE")
```



SAS and R Codes

SAS:

```
proc logistic;  
  model CAND_PTY_AFFILIATION(ref="DEM") = TTL_RECEIPTS TTL_DISB CAND_LOANS TTL_INDIV_CONTRIB / link=glogit;  
run;
```

R:

```
library(nnet)  
  
fit = multinom(party ~ receipts + CAND_LOANS + TTL_INDIV_CONTRIB,  
               data = cali)  
summary(fit)  
null = multinom(party ~ 1, data = cali)  
  
print(deviance <- deviance(null)-deviance(fit))  
  
print(p.value<- pchisq(deviance, df=12, lower.tail = FALSE))
```



SAS and R Outputs

R

```
call:
multinom(formula = party ~ receipts + CAND_LOANS + TTL_INDIV_CONTRIB,
  data = cali)
```

Coefficients:

	(Intercept)	receipts	CAND_LOANS	TTL_INDIV_CONTRIB
IND	-2.8902295	-3.840985e-04	3.049641e-04	3.853073e-04
REP	-0.2283919	-9.474754e-07	1.277966e-05	1.040096e-06

Std. Errors:

	(Intercept)	receipts	CAND_LOANS	TTL_INDIV_CONTRIB
IND	1.033757e-08	5.809849e-05	1.063057e-04	5.825450e-05
REP	4.579551e-09	4.865105e-07	4.817382e-06	5.806167e-07

Residual Deviance: 474.0105

AIC: 490.0105

```
print(deviance
L] 24.57901
```

```
print(p.value
1] 0.01694874
```




SAS and R Outputs

SAS

Parameter	CAND_PTY_AFFILIATION	DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	IND	1	-2.9924	0.4967	36.2889	<.0001
Intercept	REP	1	-0.2284	0.1242	3.3808	0.0660
TTL_RECEIPTS	IND	1	-0.00012	0.000180	0.4178	0.5181
TTL_RECEIPTS	REP	1	-9.47E-7	4.977E-7	3.6248	<u>0.0569</u>
CAND_LOANS	IND	1	0.000043	0.000244	0.0306	0.8610
CAND_LOANS	REP	1	0.000013	4.857E-6	6.9215	<u>0.0085</u>
TTL_INDIV_CONTRIB	IND	1	0.000117	0.000181	0.4196	0.5171
TTL_INDIV_CONTRIB	REP	1	1.04E-6	5.875E-7	3.1338	<u>0.0767</u>

Likelihood Ratio	23.9767	6	0.0005
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AIC	502.590	490.613
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Fitted Model, Significance, & Interpretation



Fitted model:

$$\frac{\hat{P}(IND)}{\hat{P}(DEM)} = \exp(-2.9924 - .00012(receipts) + .000244(candidate\ loans) + .000117(individual\ contributions))$$

$$\frac{\hat{P}(REP)}{\hat{P}(DEM)} = \exp(-0.2284 - 9.47E-7(receipts) + .000013(candidate\ loans) + 1.04E-6(individual\ contributions))$$

Significant Predictor - CAND_LOANS for the odds of REP vs. DEM

As candidate loans increase by \$1, the estimated odds in favor of republicans against democrats increase by $(\exp(.000013)-1)*100\% = 0.0013\%$

Prediction SAS & R

SAS

```
1  
2  
3 data pred;  
4 input TTL_RECEIPTS CAND_LOANS TTL_INDIV_CONTRIB;  
5 cards;  
6 570000 123450 300  
7 ;  
8  
9  
10 data Cali;  
11 set cali pred;  
12 run;  
13  
14  
15 proc logistic;  
16 model CAND_PTY_AFFILIATION(ref="DEM") = TTL_RECEIPTS CAND_LOANS TTL_INDIV_CONTRIB / link=glogit;  
17 output out=outdata p=poutcome;  
18 run;  
19  
20 proc print data=outdata noobs;  
21 var _level_ poutcome;  
22 run;
```

Prediction Output

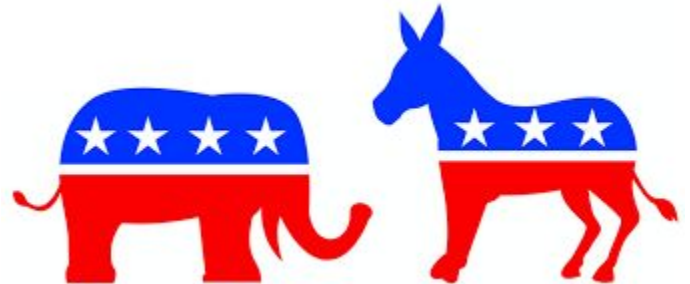
DEM	0.30802
IND	0.00000
REP	0.69198

Prediction SAS & R

R

```
> pred = predict(fit, data.frame(receipts=570000, CAND_LOANS = 123450,  
+                               TTL_INDIV_CONTRIB = 300), type = "prob")  
> print(pred)
```

	DEM	IND	REP
	3.079954e-01	3.556468e-81	6.920046e-01



Thank you!

I would like to thank

- My professor, Dr Olga
- All of my STAT 410 class
- The United States Federal Election Commission (FEC) for providing data on campaign finance contributions for all parties

