Exam Template

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Pull in Data

```
setwd("~/Documents/GitHubRepo/729_Reed_MLE_git/Exam")
#voterid <- read.csv(file = "https://raw.githubusercontent.com/Neilblund/729A/master/data/voterid.csv",
#data <- read.dta(file = "https://raw.githubusercontent.com/Neilblund/729A/master/data/voterid.csv")
#data <- read.dta13(file = "https//raw.githubusercontent.com/Neilblund/729A/master/data/voterid.csv")
#save(voterid, file = "data.RData")
load("data.RData")
#View(voterid2)</pre>
```

Introduction:

Logit

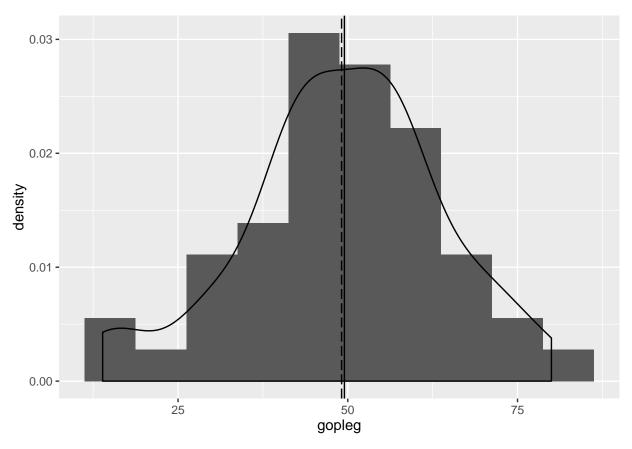
Probit

```
stargazer(voterid2, header = F)
```

Table 1:

Statistic	N	Mean	St. Dev.	Min	Max
photo	48	0.375	0.489	0	1
fraud	48	1.066	1.817	0.000	7.833
election_margin	48	15.400	10.827	4.911	49.431
gopleg	48	49.088	14.693	13.900	79.991
mean_gopleg	48	49.088	0.000	49.088	49.088
sd_gopleg	48	7.347	0.000	7.347	7.347
med_g	48	49.502	0.000	49.502	49.502

```
g <- ggplot(voterid2,aes(x=gopleg))
g + geom_histogram(aes(y=..density..),binwidth = 7.5) +
   geom_density() +
   geom_vline(xintercept = voterid2$mean_gopleg,linetype='longdash') +
   geom_vline(xintercept = voterid2$med_g)</pre>
```



```
##
## Call: glm(formula = "photo ~ fraud + election_margin + gopleg", family = binomial(link = "probit"),
##
      data = voterid2)
##
## Coefficients:
                             fraud election_margin
##
      (Intercept)
                                                              gopleg
        -2.252268
                          0.143559
                                          -0.001425
                                                          0.035747
##
## Degrees of Freedom: 47 Total (i.e. Null); 44 Residual
## Null Deviance:
                       63.51
## Residual Deviance: 54.75
                               AIC: 62.75
```

```
summary(model_1)
```

```
##
## Call:
```

```
## glm(formula = "photo ~ fraud + election_margin + gopleg", family = binomial(link = "probit"),
##
       data = voterid2)
##
## Deviance Residuals:
                 1Q
                      Median
                                   3Q
## -1.4646 -0.9162 -0.6509
                              1.0391
                                        2.4214
## Coefficients:
##
                    Estimate Std. Error z value Pr(>|z|)
                               0.828138 -2.720 0.00653 **
## (Intercept)
                   -2.252268
## fraud
                    0.143559
                               0.112251
                                          1.279 0.20093
## election_margin -0.001425
                               0.021519 -0.066 0.94721
## gopleg
                    0.035747
                               0.016206
                                          2.206 0.02739 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
       Null deviance: 63.510 on 47 degrees of freedom
## Residual deviance: 54.748 on 44 degrees of freedom
## AIC: 62.748
## Number of Fisher Scoring iterations: 5
#stargazer(model_1, type = 'text')
# generate predicted probabilities automatically
voterid2$pprob <- predict(model_1, type="response")</pre>
# generate predicted probabilities manually
voterid2$pprob_manual <- pnorm(model_1$coef['(Intercept)'] +</pre>
                                 model 1$coef['fraud']*voterid2$fraud +
                                 model 1$coef['election margin']*voterid2$election margin +
                                 model_1$coef['gopleg']*voterid2$gopleg)
# test that we did it right
voterid2$pprob_test <- voterid2$pprob - voterid2$pprob_manual</pre>
summary(voterid2$pprob_test) # should be zeros
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
##
                 0
                                 0
#######1. calculate average effect of photo using observed values----
voterid2$pprob_gopleg_upsd <- pnorm(model_1$coef['(Intercept)'] +</pre>
                                 model 1$coef['fraud']*voterid2$fraud +
                                 model_1$coef['election_margin']*voterid2$election_margin +
                                 model_1$coef['gopleg']*gopleg_obs_high)
summary(voterid2$pprob_gopleg_upsd)
     Min. 1st Qu. Median
                              Mean 3rd Qu.
## 0.3801 0.4035 0.4194 0.4574 0.4653 0.8036
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.2031 0.2208 0.2331 0.2695 0.2702 0.6290
```

voterid2\$pprob_effect <- voterid2\$pprob_gopleg_upsd - voterid2\$pprob_gopleg_downsd summary(voterid2\$pprob_effect)

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.1746 0.1825 0.1855 0.1879 0.1921 0.2071
```

Summary: Interpreting the Coefficients, include the AIC

```
stargazer(model_1, header = F)
```

Table 2:

	Dependent variable:	
	NA	
fraud	0.144	
	(0.112)	
election_margin	-0.001	
_ 0	(0.022)	
gopleg	0.036**	
	(0.016)	
Constant	-2.252***	
	(0.828)	
Observations	48	
Log Likelihood	-27.374	
Akaike Inf. Crit.	62.748	
Note:	*p<0.1; **p<0.05; ***p<0.05	