

# STORYTELLING MODEL INTERPRETATION

1 NOVEMBER 2019  
BADM HACKATHON TRAINING DAY



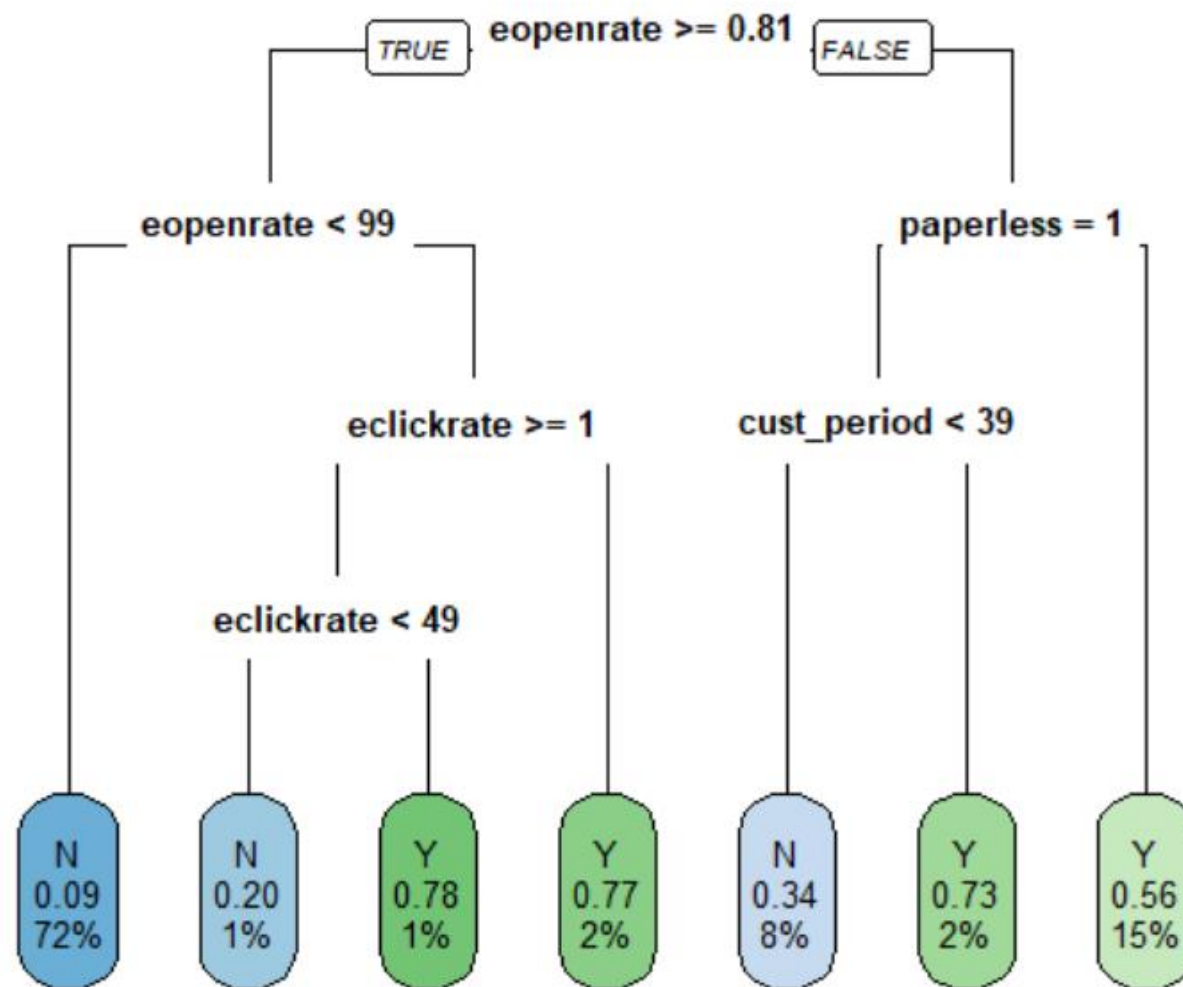
SFU  
**BEEDIE**  
SCHOOL OF BUSINESS



# MODEL OUTPUTS

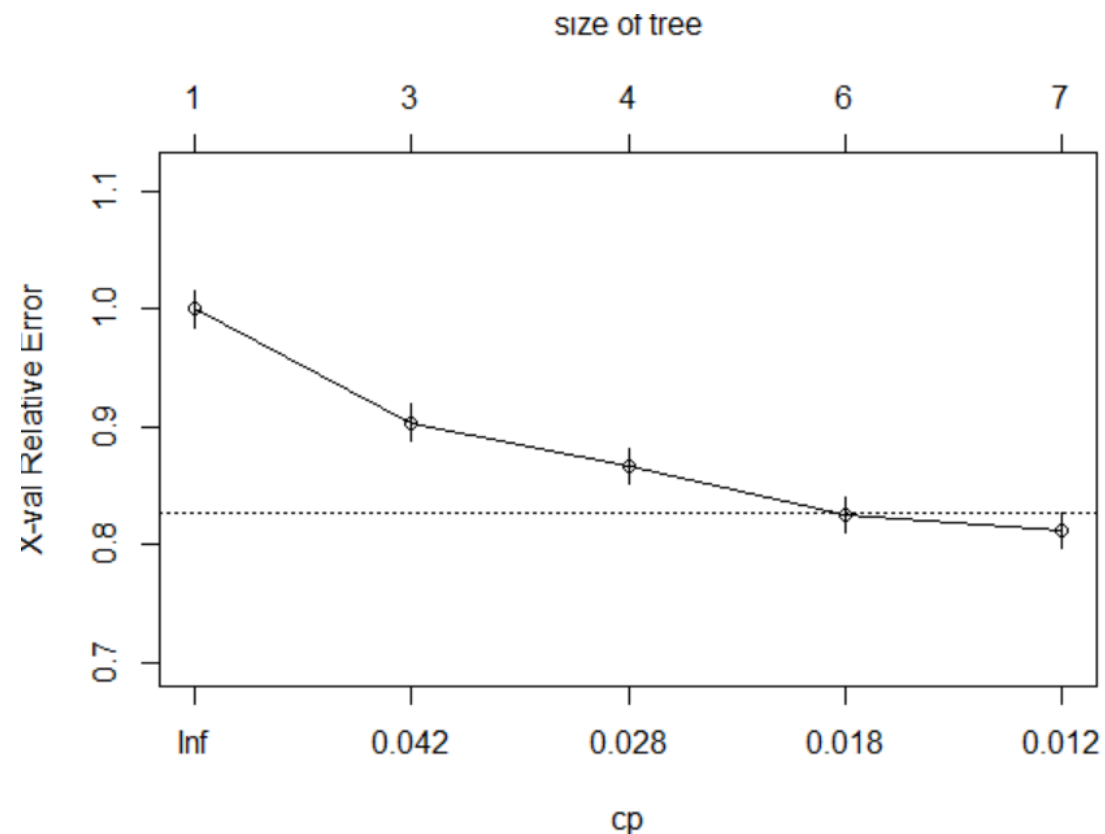
# DECISION TREE

Model1.RPart



# DECISION TREE

0.01  
Complexity  
Parameter



variables actually used in tree construction:  
 [1] cust\_period eclickrate eopenrate paperless  
 Root node error: 3147/15364 = 0.20483 n= 15364  
 CP nsplit rel error xerror xstd 1 0.048141 0  
 1.00000 1.00000 0.015896 2 0.037496 2 0.90372  
 0.90372 0.015297 3 0.020813 3 0.86622 0.86718  
 0.015054 4 0.015570 5 0.82459 0.82555 0.014764 5  
 0.010000 6 0.80902 0.81220 0.014668

# LOGISTIC REGRESSION

## Summary

```
Call:
glm(formula = lost ~ eopenrate + eclickrate + avgorder + ordfreq +
    paperless + refill + doorstep + favday + city + cust_period,
    family = binomial(logit), data = filter(Retention2017, Sample ==
    "Estimation"))
```

```
Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.1665  -0.7025  -0.5891  -0.4095   2.7587
```

```
Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept)  -1.1791340  0.1385997  -8.507  < 2e-16 ***
eopenrate     -0.0030877  0.0008689  -3.554  0.00038 ***
eclickrate     0.0052301  0.0025311   2.066  0.03880 *
avgorder      -0.0017974  0.0005858  -3.068  0.00215 **
ordfreq       -0.2589222  0.2093322  -1.237  0.21613
paperless     -0.7461929  0.0463274 -16.107  < 2e-16 ***
refill        -0.7789828  0.0988066  -7.884  3.17e-15 ***
doorstep      -0.6818040  0.1682071  -4.053  5.05e-05 ***
favdayMonday  -0.0289822  0.0670705  -0.432  0.66566
favdaySaturday 0.0942395  0.1335878   0.705  0.48053
favdaySunday   0.4632740  0.1454314   3.186  0.00144 **
favdayThursday 0.0455955  0.0703201   0.648  0.51673
favdayTuesday  0.0959347  0.0664393   1.444  0.14875
favdayWednesday 0.0939152  0.0707280   1.328  0.18423
cityCHO        0.7440737  0.1228154   6.058  1.37e-09 ***
cityDCX        0.3260985  0.1185809   2.750  0.00596 **
cityRIC        0.4890922  0.1213393   4.031  5.56e-05 ***
cust_period    -0.0005735  0.0001011  -5.674  1.39e-08 ***
---
```

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

```
(Dispersion parameter for binomial family taken to be 1)
```

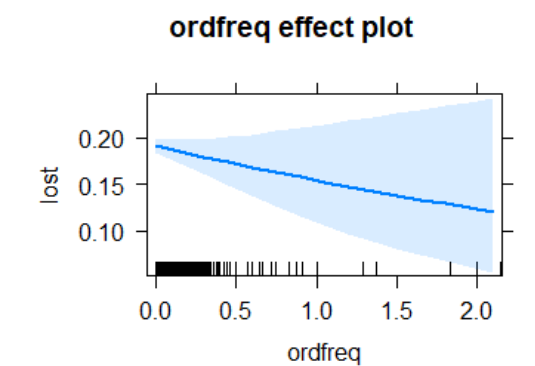
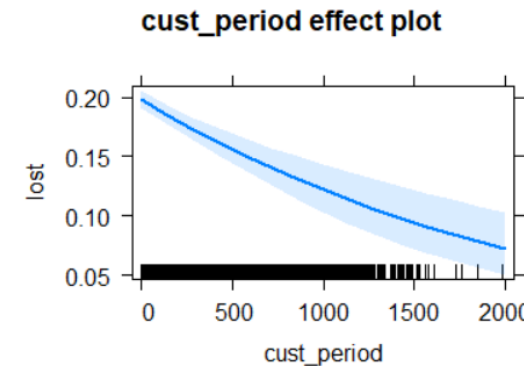
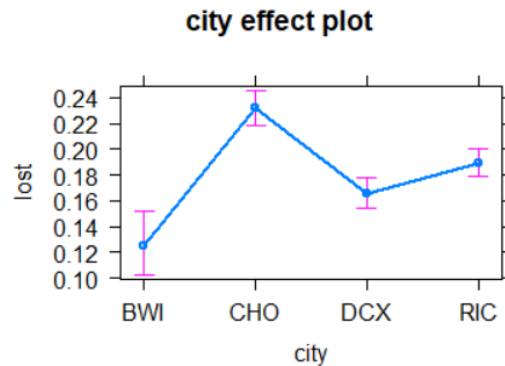
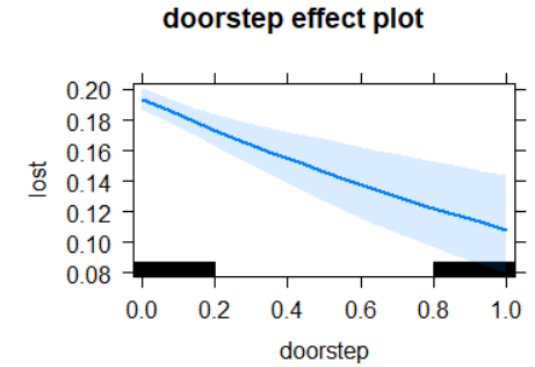
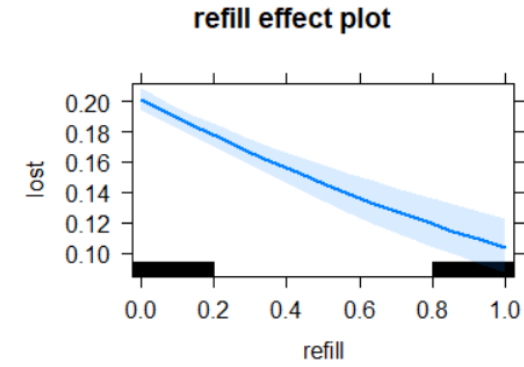
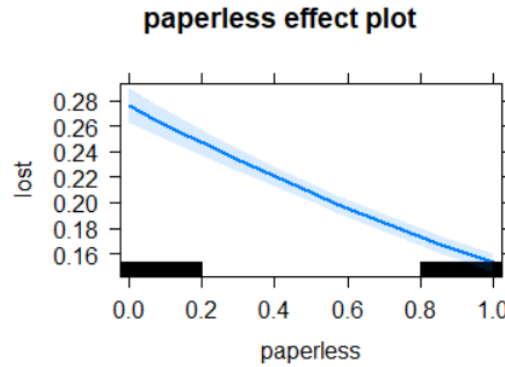
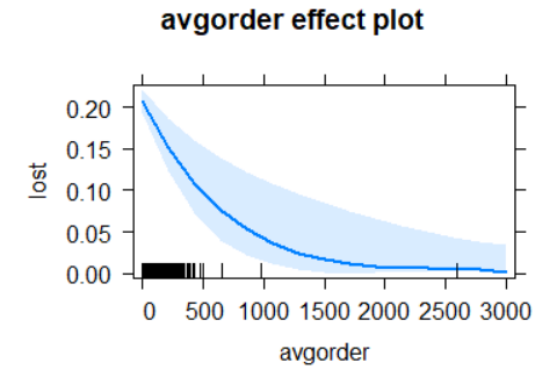
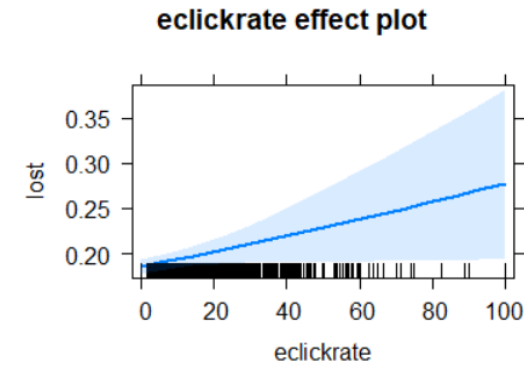
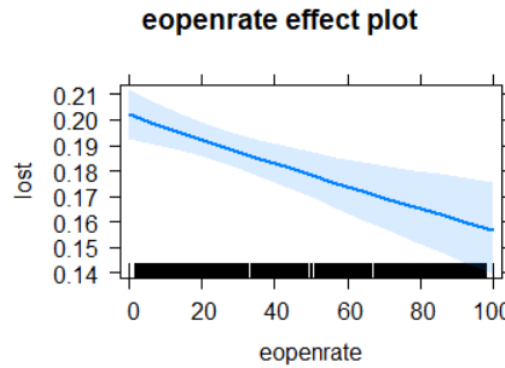
```
Null deviance: 15580  on 15363  degrees of freedom
Residual deviance: 14838  on 15346  degrees of freedom
AIC: 14874
```

```
Number of Fisher Scoring iterations: 5
```



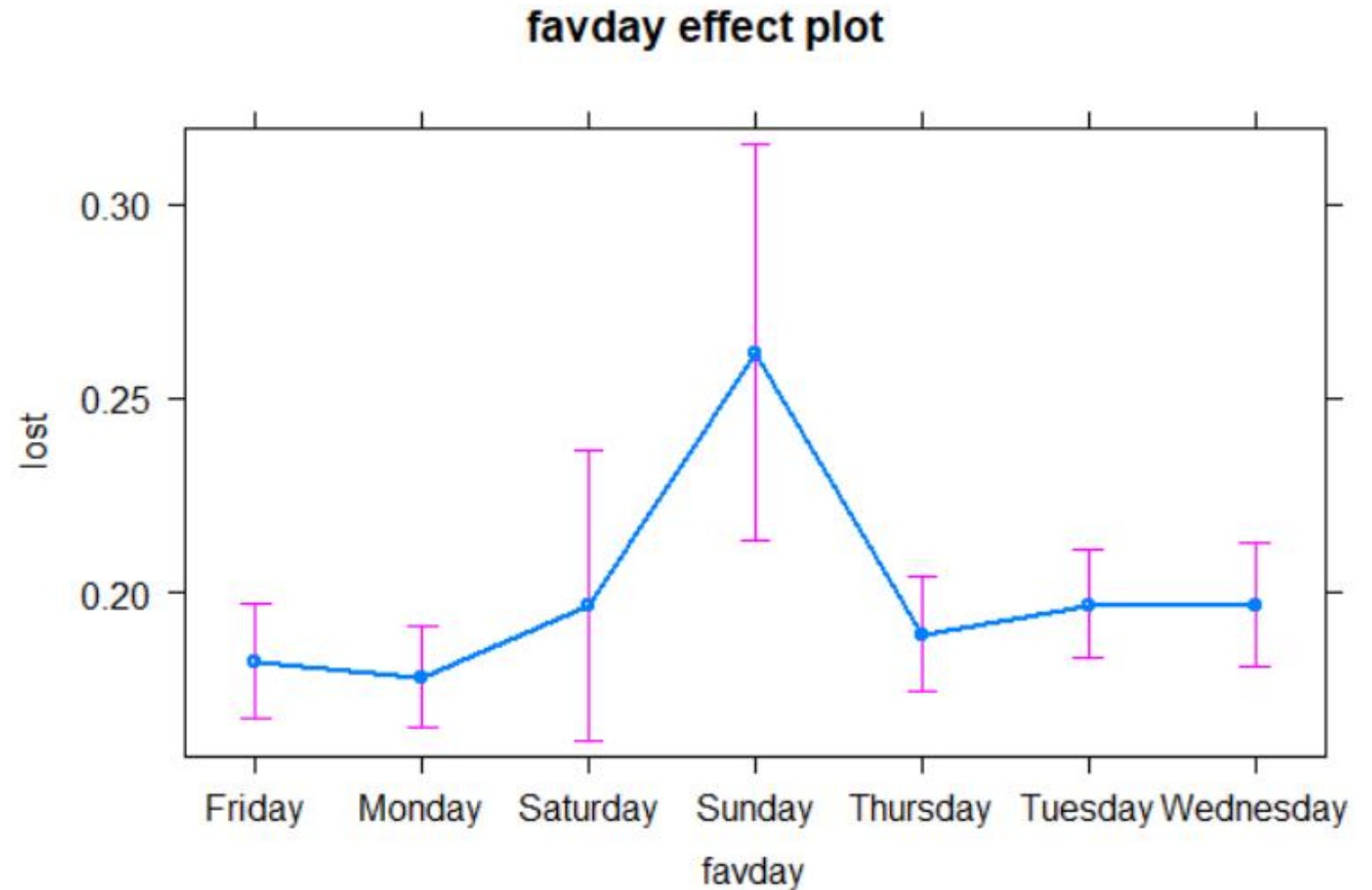
# LOGISTIC REGRESSION

## Effect Plots



# LOGISTIC REGRESSION

Effect Plots  
Cont.



# STEPWISE REGRESSION

## Summary

```
Call:
glm(formula = lost ~ eopenrate + eclickrate + avgorder + paperless +
     refill + doorstep + favday + city + cust_period, family = binomial(logit),
     data = filter(Retention2017, Sample == "Estimation"))
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.1606	-0.7026	-0.5893	-0.4107	2.7597

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )	
(Intercept)	-1.1877134	0.1384274	-8.580	< 2e-16	***
eopenrate	-0.0030927	0.0008690	-3.559	0.000373	***
eclickrate	0.0051114	0.0025310	2.020	0.043430	*
avgorder	-0.0018368	0.0005860	-3.134	0.001722	**
paperless	-0.7436690	0.0462733	-16.071	< 2e-16	***
refill	-0.7846520	0.0987058	-7.949	1.87e-15	***
doorstep	-0.6890675	0.1680878	-4.099	4.14e-05	***
favdayMonday	-0.0293886	0.0670663	-0.438	0.661240	
favdaySaturday	0.0955010	0.1335632	0.715	0.474594	
favdaySunday	0.4579039	0.1453359	3.151	0.001629	**
favdayThursday	0.0458790	0.0703158	0.652	0.514097	
favdayTuesday	0.0951621	0.0664318	1.432	0.152007	
favdayWednesday	0.0922542	0.0707152	1.305	0.192033	
cityCHO	0.7453465	0.1227835	6.070	1.28e-09	***
cityDCX	0.3282712	0.1185503	2.769	0.005622	**
cityRIC	0.4916868	0.1212979	4.054	5.04e-05	***
cust_period	-0.0005723	0.0001011	-5.662	1.50e-08	***

---

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

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 15580 on 15363 degrees of freedom  
Residual deviance: 14839 on 15347 degrees of freedom  
AIC: 14873

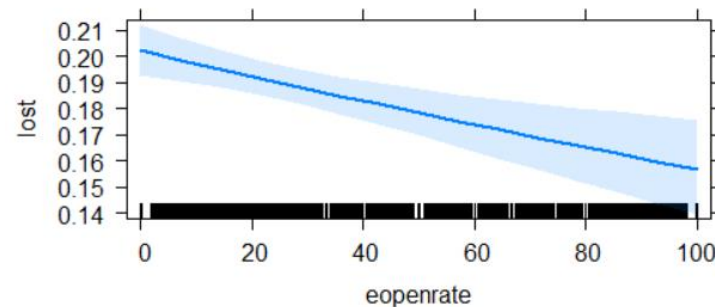
Number of Fisher Scoring iterations: 5



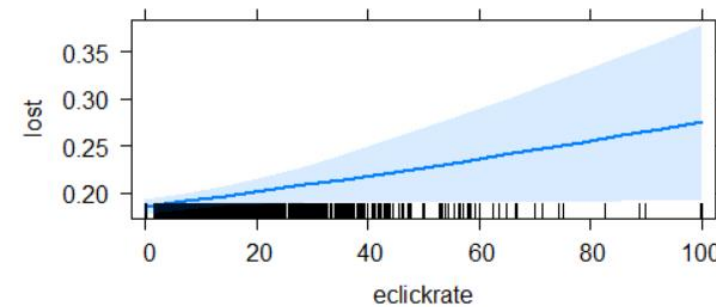
# STEPWISE REGRESSION

## Effect Plots

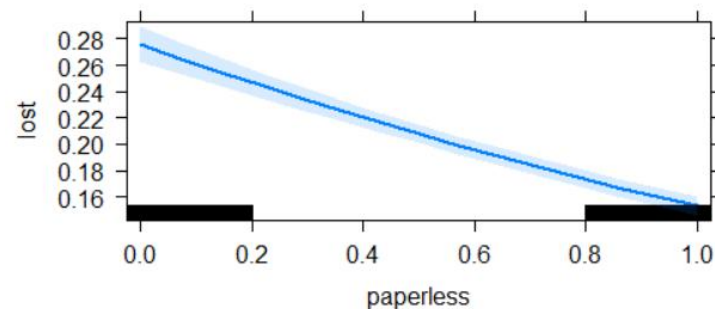
eopenrate effect plot



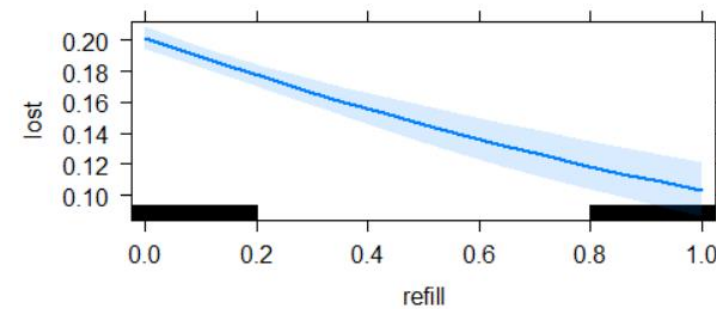
eclickrate effect plot



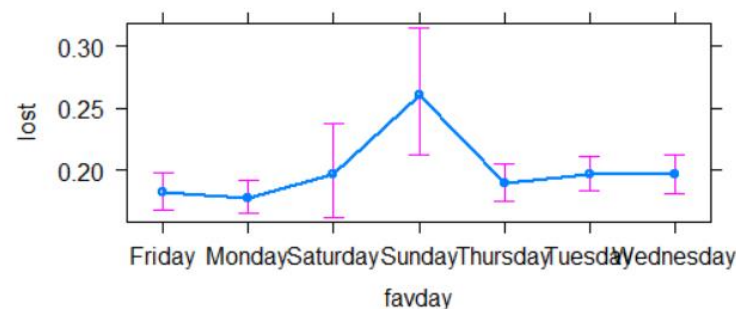
paperless effect plot



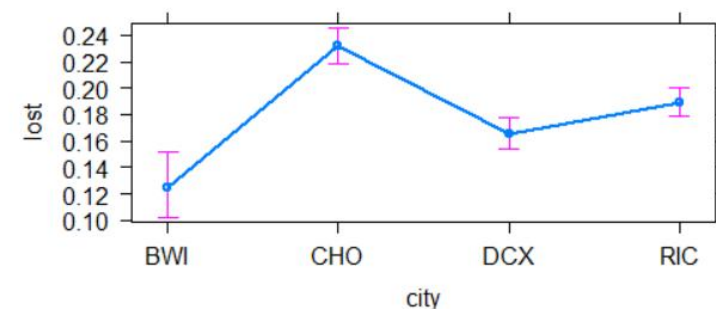
refill effect plot



favday effect plot



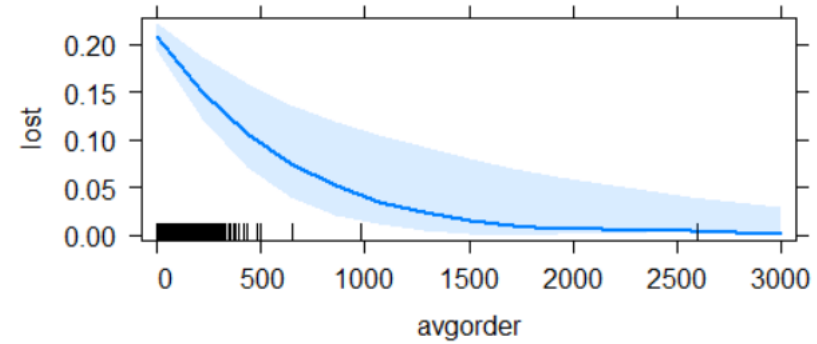
city effect plot



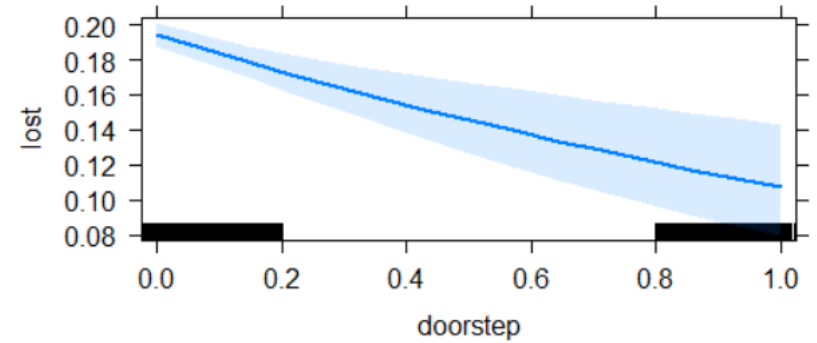
# STEPWISE REGRESSION

## Effect Plots Cont.

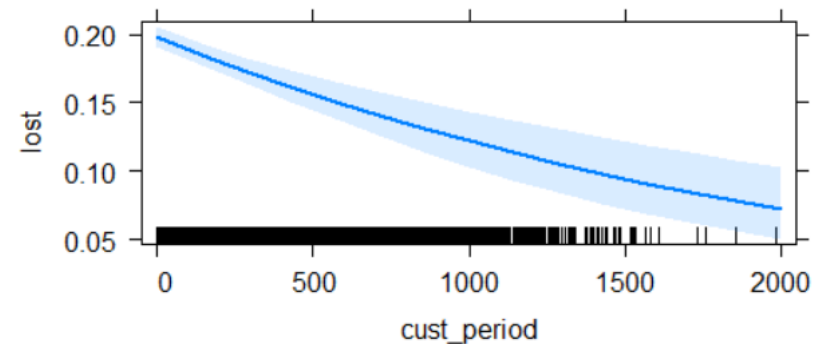
avgorder effect plot



doorstep effect plot



cust\_period effect plot



# RANDOM FOREST

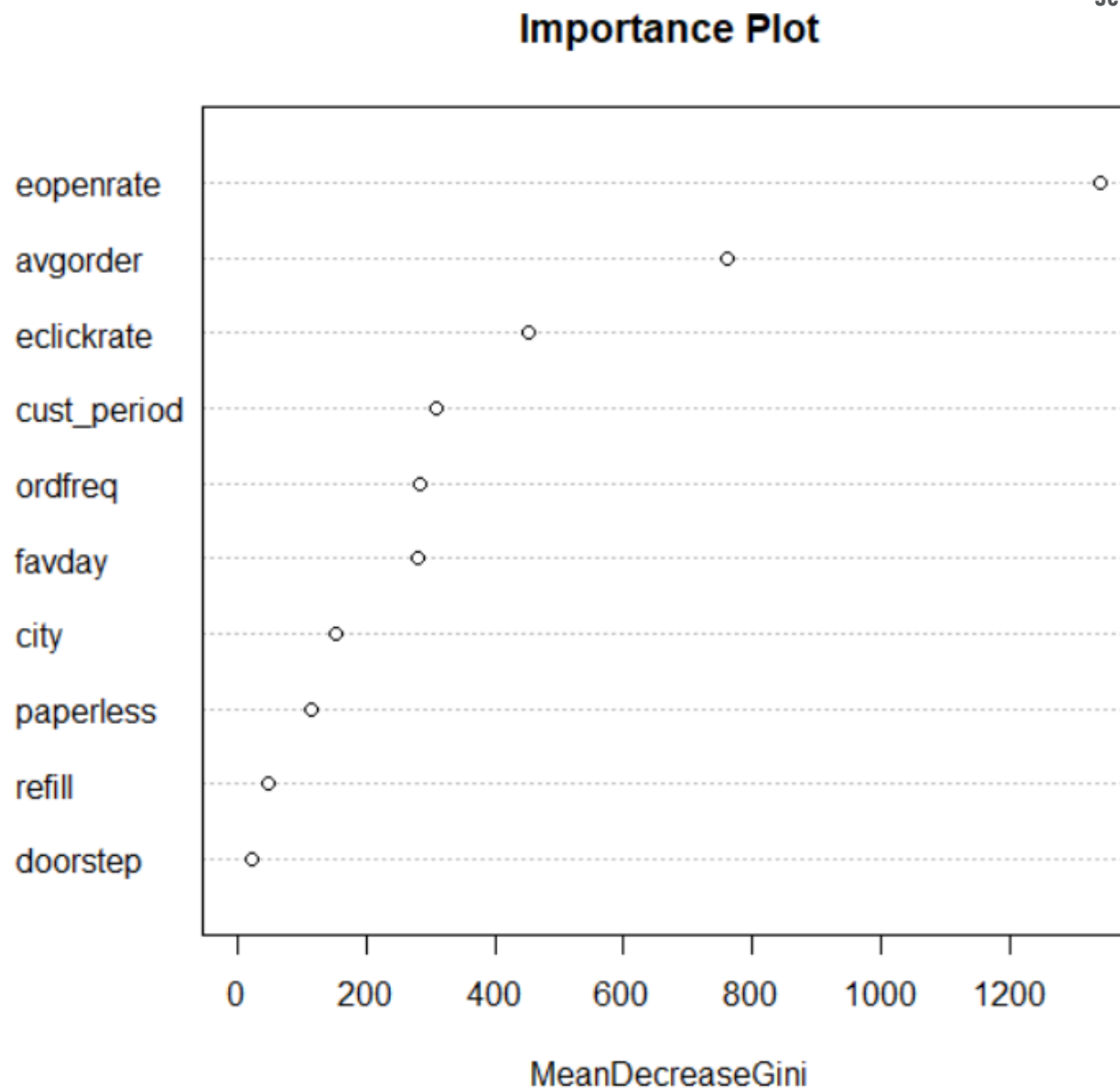
## Summary

```
Call:
  randomForest(formula = lost ~ eopenrate + eclickrate + avgorder +      ordfreq + paperless + refill + door
step + favday + city +      cust_period, data = filter(Retention2017, Sample == "Estimation"),      importa
nce = TRUE, ntree = 500, mtry = 4)
      Type of random forest: classification
      Number of trees: 500
No. of variables tried at each split: 4

      OOB estimate of  error rate: 17.08%
Confusion matrix:
      N      Y class.error
N 11184 1033  0.08455431
Y  1591 1556  0.50556085
```

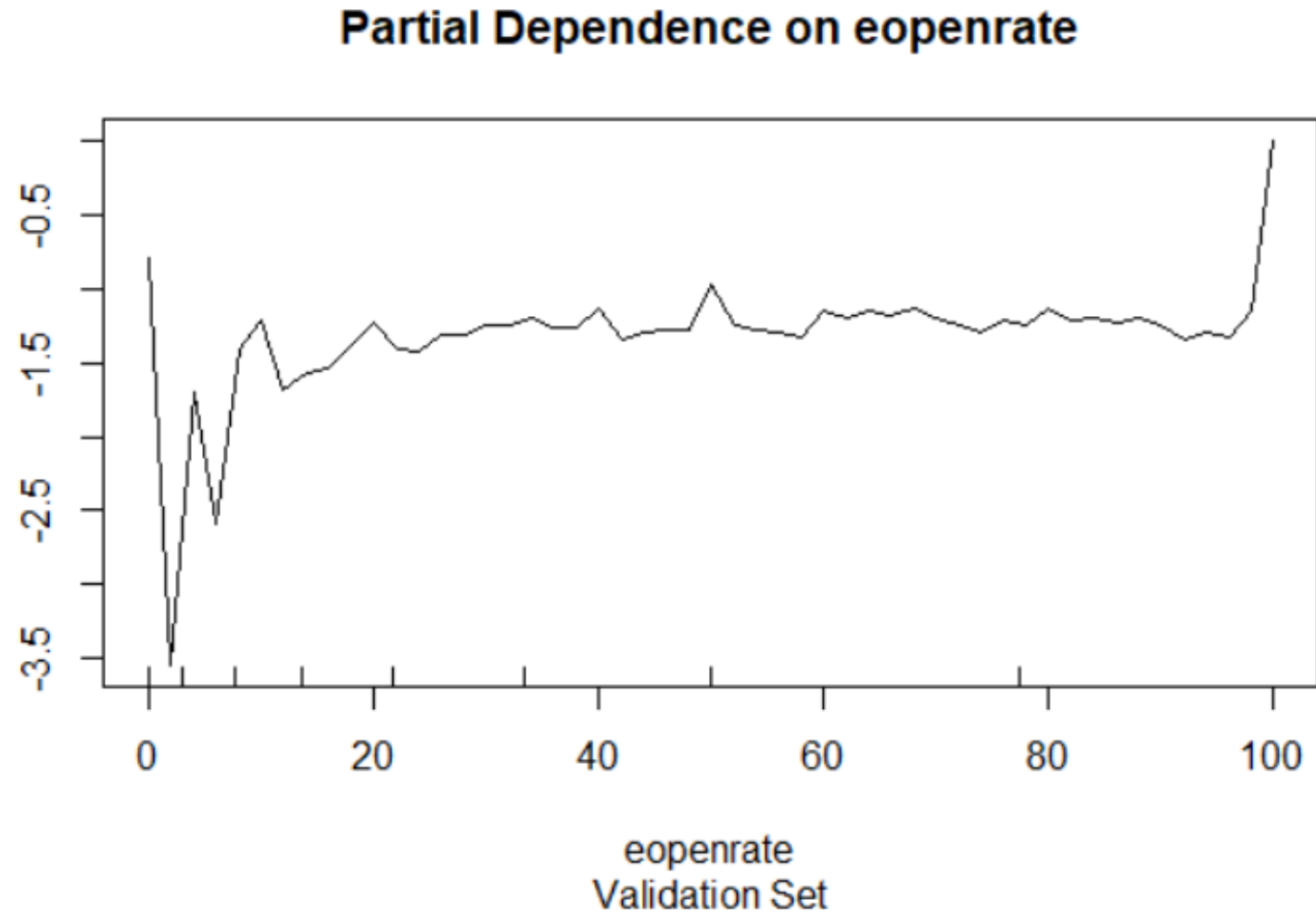
# RANDOM FOREST

## Importance Plot



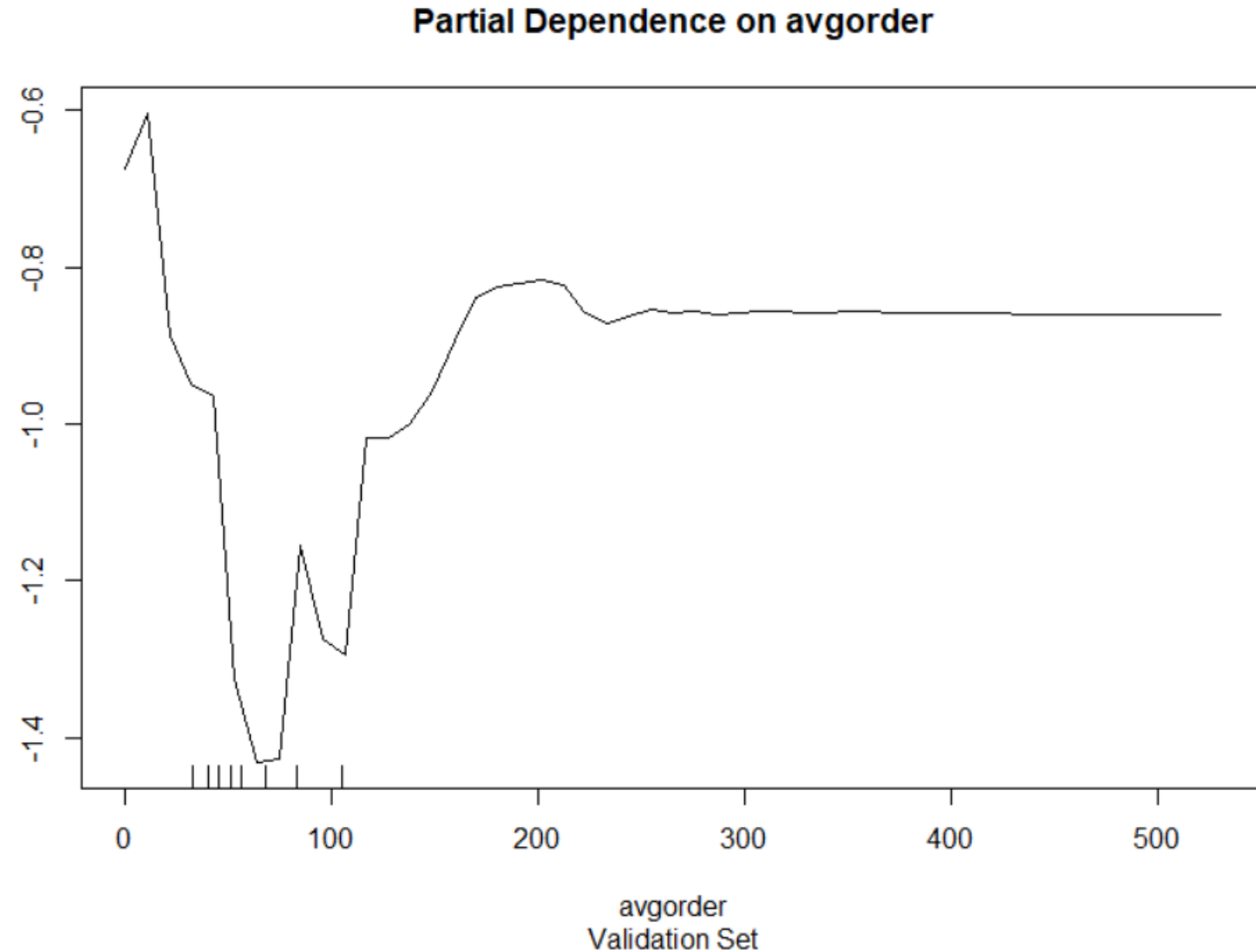
# RANDOM FOREST

Partial  
Dependence  
Plot  
eopenrate



# RANDOM FOREST

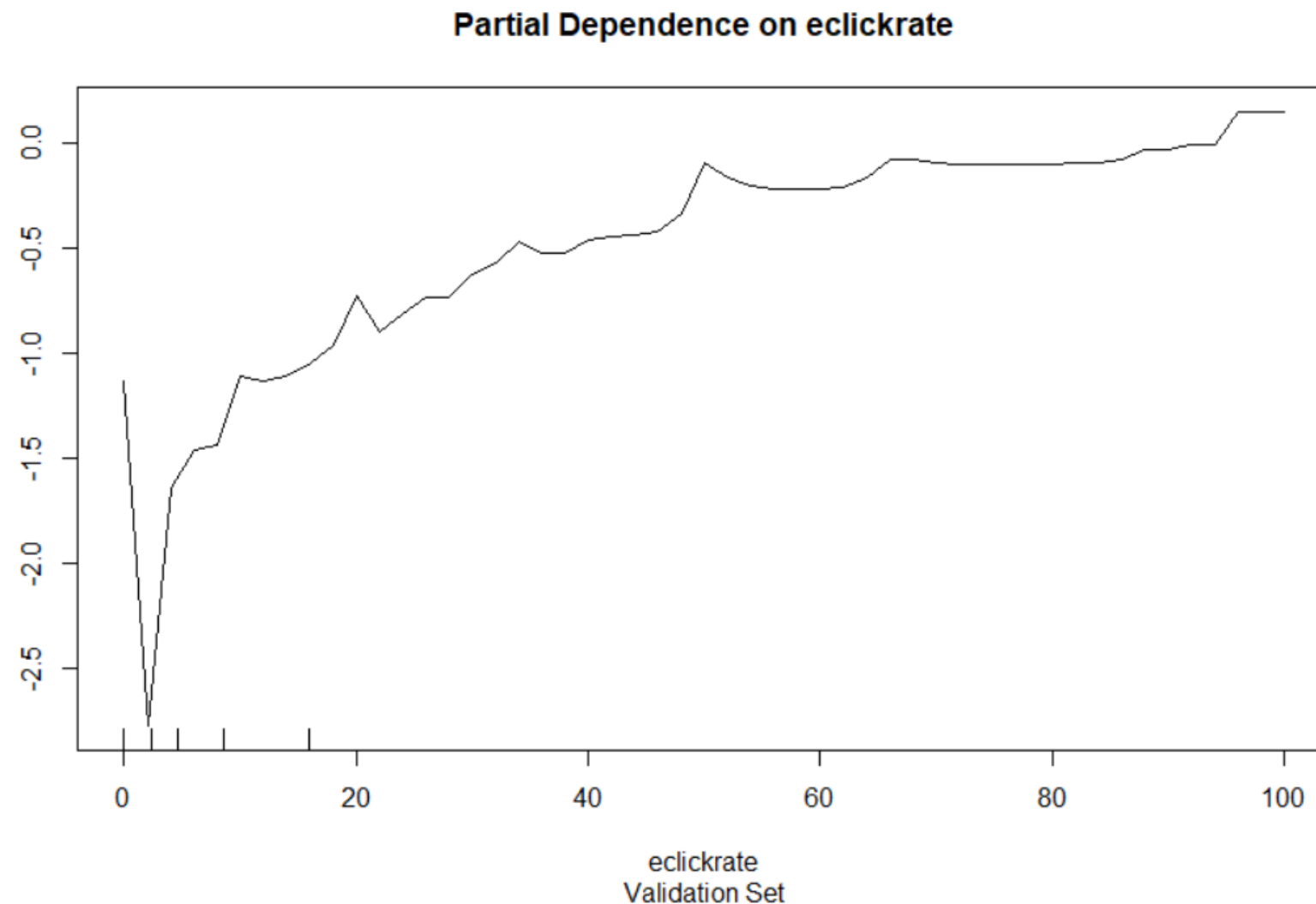
Partial  
Dependence  
Plot  
avgorder





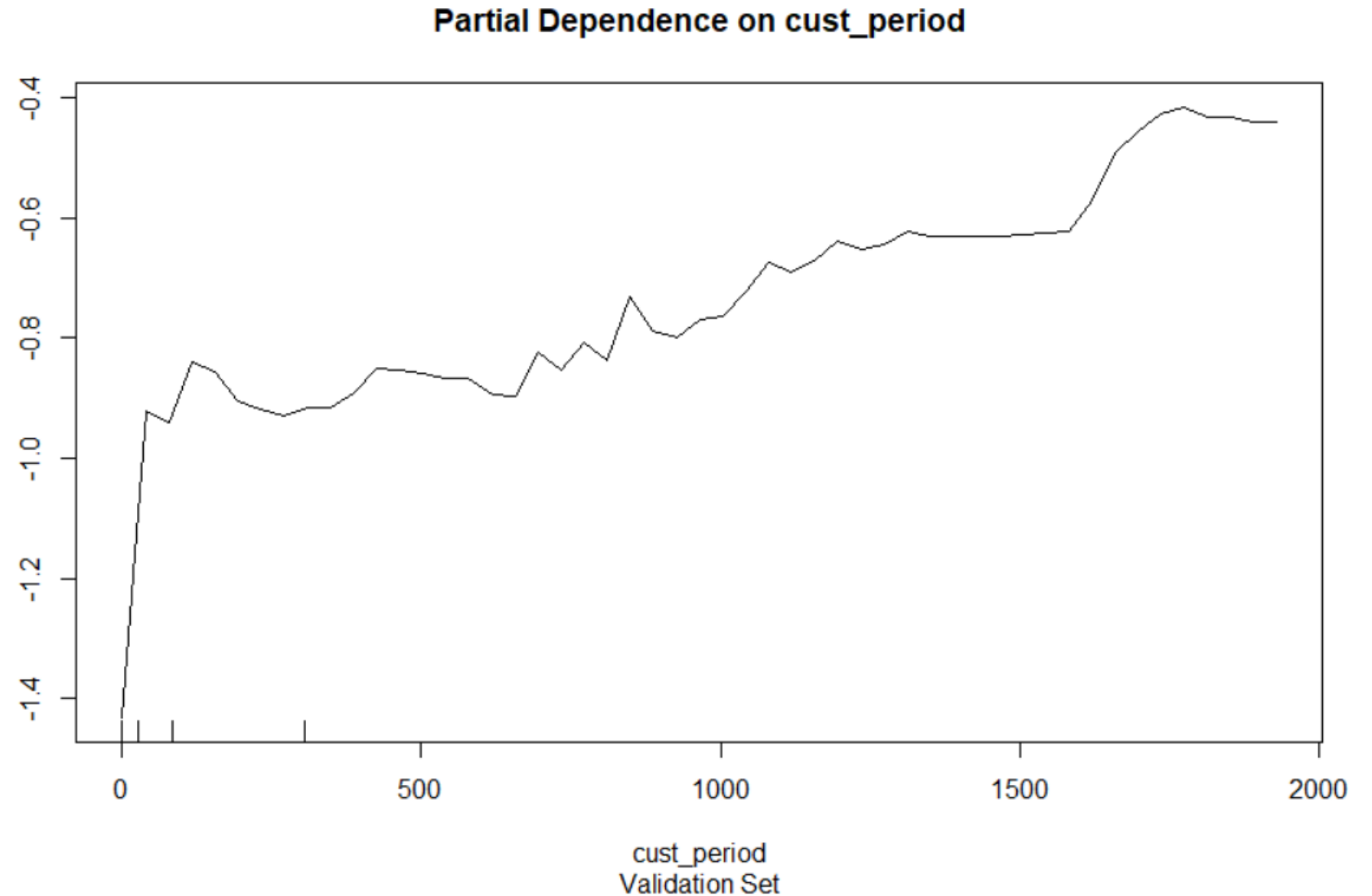
# RANDOM FOREST

Partial  
Dependence  
Plot  
eclickrate



# RANDOM FOREST

Partial  
Dependence  
Plot  
cust\_period



SFU

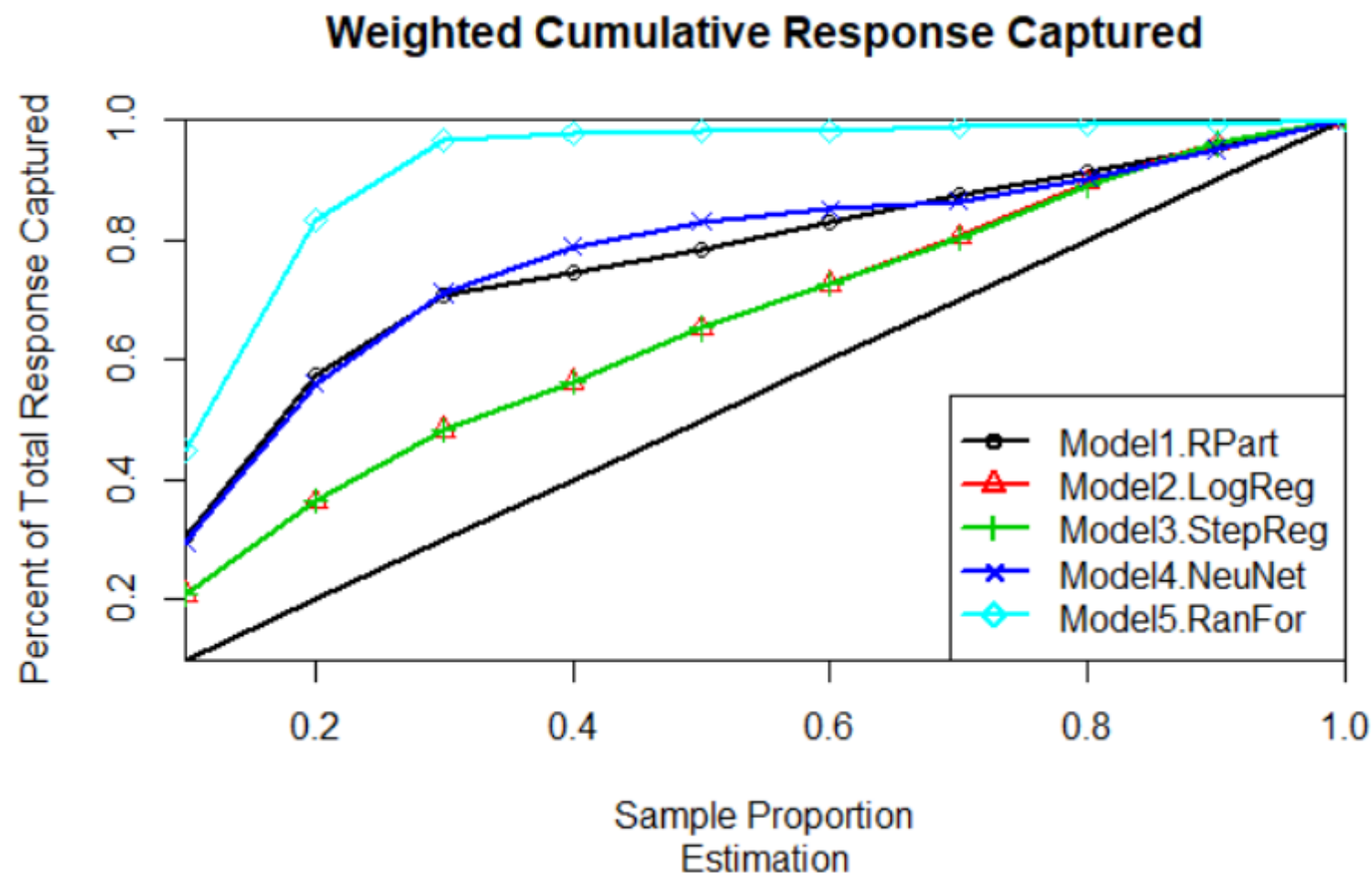
**BEEDIE**  
SCHOOL OF BUSINESS



# LIFT CHARTS

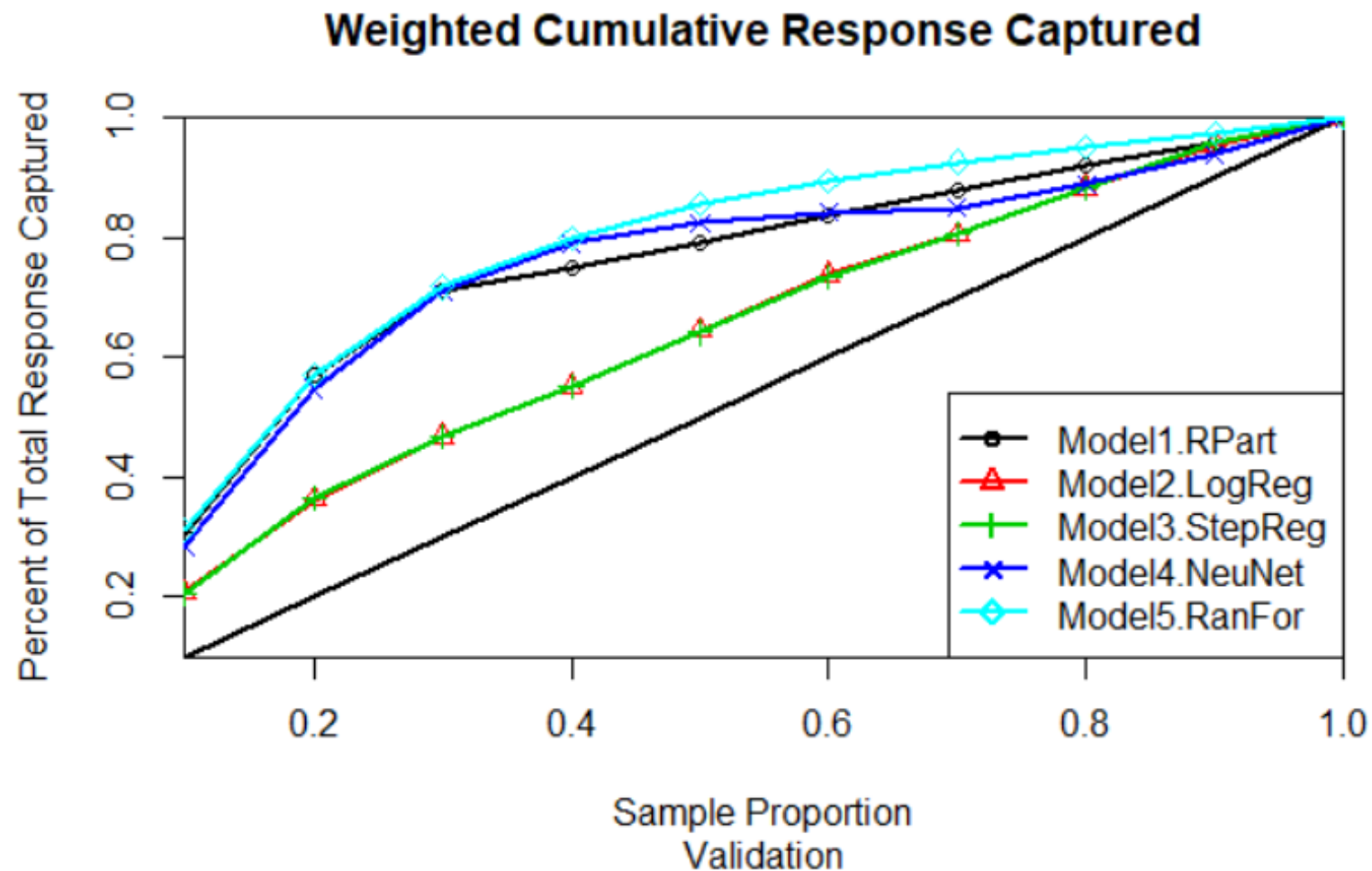
# CUMULATIVE LIFT

Estimation



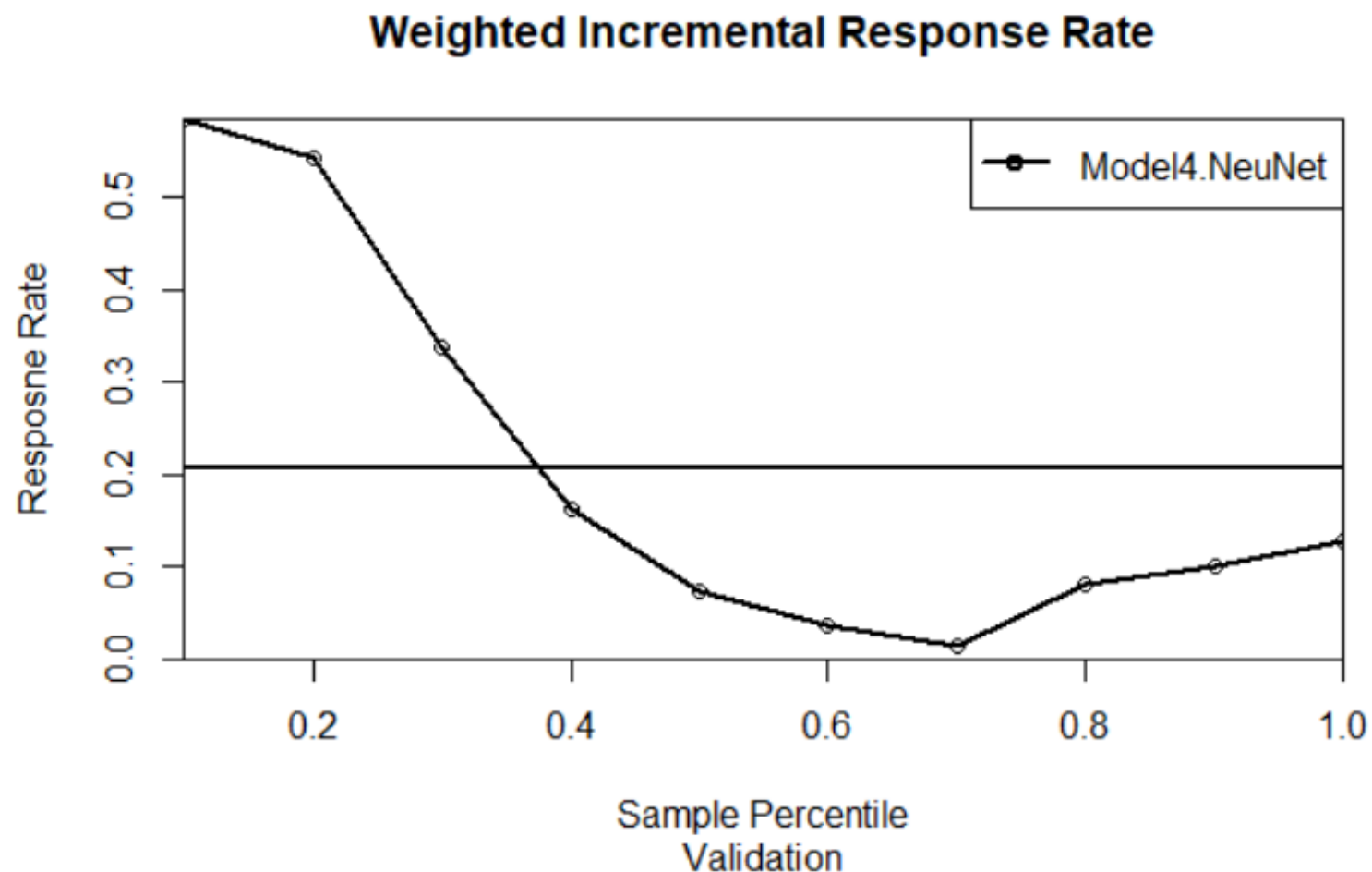
# CUMULATIVE LIFT

Validation



# INCREMENTAL LIFT

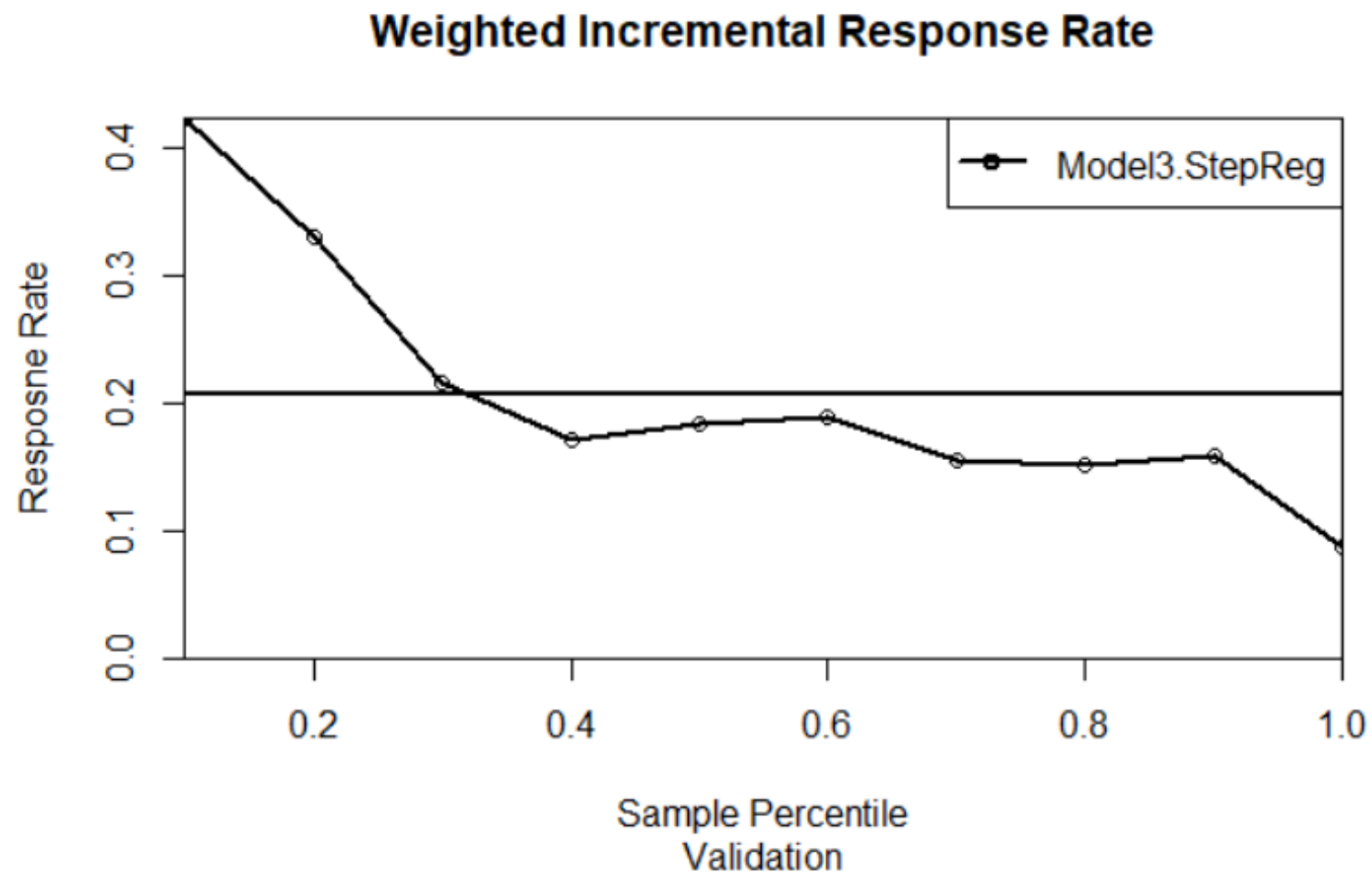
Neural  
Network  
Validation





# INCREMENTAL LIFT

Stepwise  
Regression  
Validation





# **PUTTING IT ALL TOGETHER**

# **FINANCIAL IMPACT**

# FINANCIAL IMPACT

Loss per  
order if churn

custid	avgorder	ordfreq	cust_period	Sample	score	Potential Loss
NF85M6	40.02	0		0 Holdout	0.98 \$	40.02
KL7E2H	40.02	0		0 Holdout	0.98 \$	40.02
C56P7B	40.02	0		0 Holdout	0.98 \$	40.02
9JMNAX	40.02	0		0 Holdout	0.98 \$	40.02
2C8YQT	40.02	0		0 Holdout	0.98 \$	40.02
7XP45K	14.63	0		0 Holdout	0.974 \$	14.63
P5MSYK	74.66	0.018480493		487 Holdout	0.864 \$	671.94
46V8DD	35.29	0.004413063		1133 Holdout	0.862 \$	176.45
VC3DLQ	15.95	0.578947368		19 Holdout	0.86 \$	175.45
JS7YEU	103.28	0.012820513		546 Holdout	0.858 \$	722.96
QXZWPL	203.55	0.032085561		187 Holdout	0.858 \$	1,221.30
NW87UP	58.38	0.081632653		49 Holdout	0.848 \$	233.52
8NAKCE	117.66	0.011644833		687 Holdout	0.844 \$	941.28
SAC5P6	93.87	0.022288262		673 Holdout	0.844 \$	1,408.05
FAS9W3	25.63	0.013793103		145 Holdout	0.844 \$	51.26
HTSXKU	88.04	0.033613445		119 Holdout	0.842 \$	352.16
					Total	\$ 6,169.10

Potential Loss = Average Order x Order Frequency x Customer Period  
 Or if order frequency is 0  
 Potential Loss = Average Order