

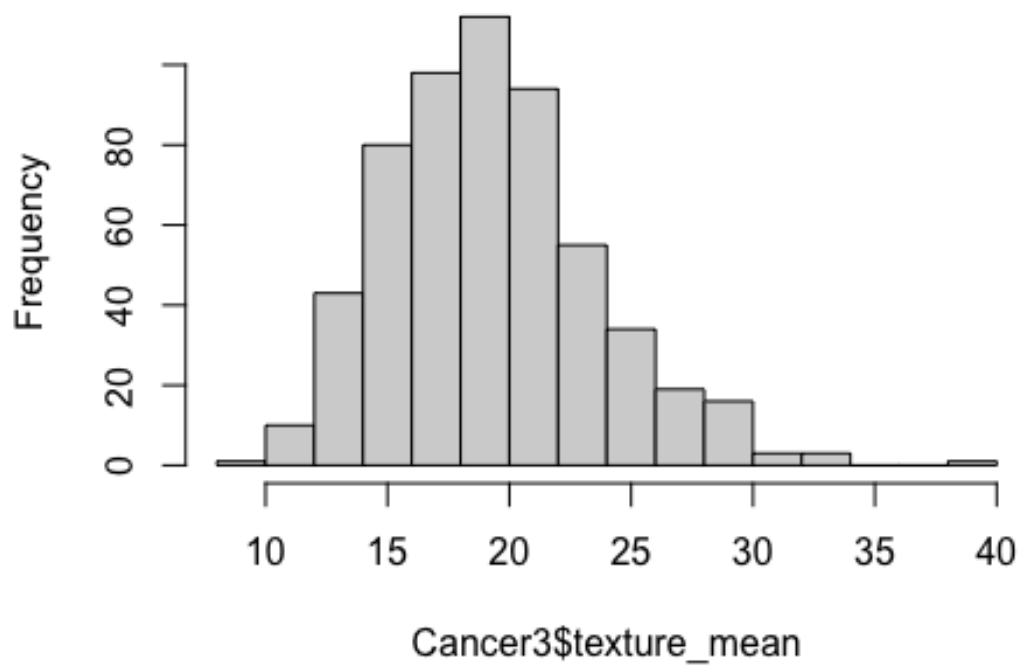
ML Final project Linear and Logistics

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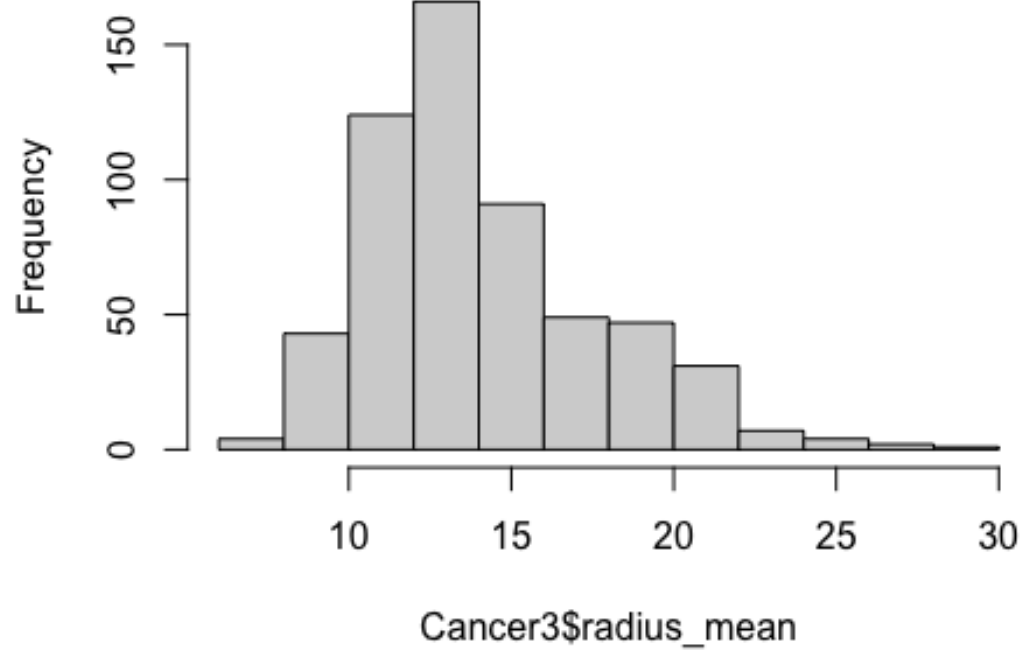
```
options(scipen=999)
set.seed(1)
Cancer3 <- read.csv("/Users/nawwaf/Desktop/Kent/Kent Master_s/Machine
Learning/Final Project/data.csv")
hist(Cancer3$texture_mean)
```

Histogram of Cancer3\$texture_mean



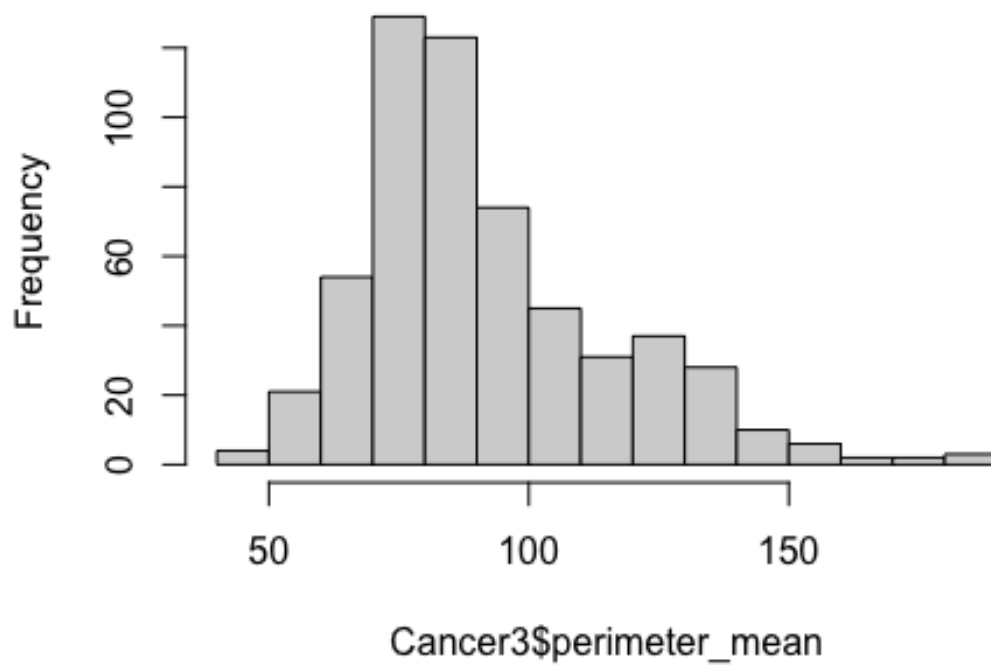
```
hist(Cancer3$radius_mean)
```

Histogram of Cancer3\$radius_mean



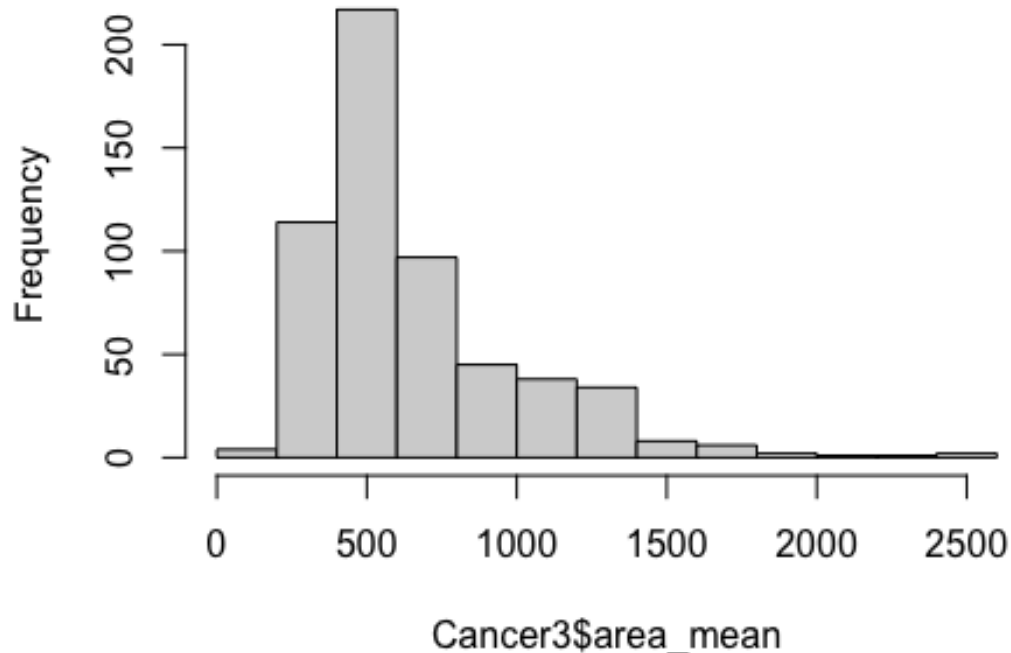
```
hist(Cancer3$perimeter_mean)
```

Histogram of Cancer3\$perimeter_mean



```
hist(Cancer3$area_mean)
```

Histogram of Cancer3\$area_mean



```
mean(Cancer3$concavity_worst)
```

```
## [1] 0.2721885
```

```
summary(Cancer3)
```

```
##      id      diagnosis      radius_mean      texture_mean
## Min.   : 8670 Length:569 Min.   : 6.981 Min.   : 9.71
## 1st Qu.: 869218 Class :character 1st Qu.:11.700 1st Qu.:16.17
## Median : 906024 Mode  :character Median :13.370 Median :18.84
## Mean   : 30371831 Mean   :14.127 Mean   :19.29
## 3rd Qu.: 8813129 3rd Qu.:15.780 3rd Qu.:21.80
## Max.   :911320502 Max.   :28.110 Max.   :39.28
## perimeter_mean      area_mean      smoothness_mean      compactness_mean
## Min.   : 43.79 Min.   : 143.5 Min.   :0.05263 Min.   :0.01938
## 1st Qu.: 75.17 1st Qu.: 420.3 1st Qu.:0.08637 1st Qu.:0.06492
## Median : 86.24 Median : 551.1 Median :0.09587 Median :0.09263
## Mean   : 91.97 Mean   : 654.9 Mean   :0.09636 Mean   :0.10434
## 3rd Qu.:104.10 3rd Qu.: 782.7 3rd Qu.:0.10530 3rd Qu.:0.13040
## Max.   :188.50 Max.   :2501.0 Max.   :0.16340 Max.   :0.34540
## concavity_mean      concave.points_mean      symmetry_mean
## fractal_dimension_mean
## Min.   :0.00000 Min.   :0.00000 Min.   :0.1060 Min.   :0.04996
## 1st Qu.:0.02956 1st Qu.:0.02031 1st Qu.:0.1619 1st Qu.:0.05770
```

```

## Median :0.06154 Median :0.03350 Median :0.1792 Median :0.06154
## Mean :0.08880 Mean :0.04892 Mean :0.1812 Mean :0.06280
## 3rd Qu.:0.13070 3rd Qu.:0.07400 3rd Qu.:0.1957 3rd Qu.:0.06612
## Max. :0.42680 Max. :0.20120 Max. :0.3040 Max. :0.09744
## radius_se texture_se perimeter_se area_se
## Min. :0.1115 Min. :0.3602 Min. : 0.757 Min. : 6.802
## 1st Qu.:0.2324 1st Qu.:0.8339 1st Qu.: 1.606 1st Qu.: 17.850
## Median :0.3242 Median :1.1080 Median : 2.287 Median : 24.530
## Mean :0.4052 Mean :1.2169 Mean : 2.866 Mean : 40.337
## 3rd Qu.:0.4789 3rd Qu.:1.4740 3rd Qu.: 3.357 3rd Qu.: 45.190
## Max. :2.8730 Max. :4.8850 Max. :21.980 Max. :542.200
## smoothness_se compactness_se concavity_se concave.points_se
## Min. :0.001713 Min. :0.002252 Min. :0.00000 Min. :0.000000
## 1st Qu.:0.005169 1st Qu.:0.013080 1st Qu.:0.01509 1st Qu.:0.007638
## Median :0.006380 Median :0.020450 Median :0.02589 Median :0.010930
## Mean :0.007041 Mean :0.025478 Mean :0.03189 Mean :0.011796
## 3rd Qu.:0.008146 3rd Qu.:0.032450 3rd Qu.:0.04205 3rd Qu.:0.014710
## Max. :0.031130 Max. :0.135400 Max. :0.39600 Max. :0.052790
## symmetry_se fractal_dimension_se radius_worst texture_worst
## Min. :0.007882 Min. :0.0008948 Min. : 7.93 Min. :12.02
## 1st Qu.:0.015160 1st Qu.:0.0022480 1st Qu.:13.01 1st Qu.:21.08
## Median :0.018730 Median :0.0031870 Median :14.97 Median :25.41
## Mean :0.020542 Mean :0.0037949 Mean :16.27 Mean :25.68
## 3rd Qu.:0.023480 3rd Qu.:0.0045580 3rd Qu.:18.79 3rd Qu.:29.72
## Max. :0.078950 Max. :0.0298400 Max. :36.04 Max. :49.54
## perimeter_worst area_worst smoothness_worst compactness_worst
## Min. : 50.41 Min. : 185.2 Min. :0.07117 Min. :0.02729
## 1st Qu.: 84.11 1st Qu.: 515.3 1st Qu.:0.11660 1st Qu.:0.14720
## Median : 97.66 Median : 686.5 Median :0.13130 Median :0.21190
## Mean :107.26 Mean : 880.6 Mean :0.13237 Mean :0.25427
## 3rd Qu.:125.40 3rd Qu.:1084.0 3rd Qu.:0.14600 3rd Qu.:0.33910
## Max. :251.20 Max. :4254.0 Max. :0.22260 Max. :1.05800
## concavity_worst concave.points_worst symmetry_worst
fractal_dimension_worst
## Min. :0.0000 Min. :0.00000 Min. :0.1565 Min. :0.05504
## 1st Qu.:0.1145 1st Qu.:0.06493 1st Qu.:0.2504 1st Qu.:0.07146
## Median :0.2267 Median :0.09993 Median :0.2822 Median :0.08004
## Mean :0.2722 Mean :0.11461 Mean :0.2901 Mean :0.08395
## 3rd Qu.:0.3829 3rd Qu.:0.16140 3rd Qu.:0.3179 3rd Qu.:0.09208
## Max. :1.2520 Max. :0.29100 Max. :0.6638 Max. :0.20750
## X
## Mode:logical
## NA's:569
##
##
##
##

```

```

#converting the categorical variable to factor
Cancer3$diagnosis = as.factor(Cancer3$diagnosis)

```

```
Cancer3Final = Cancer3[,c(-1,-13,-14,-15,-16,-17,-18,-19,-20,-21,-22,-23,-
-24,-25,-26,-27,-28,-29,-30,-31,-32,-33,-34)] #remove id and X
summary(Cancer3Final)
```

##	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean
##	B:357	Min. : 6.981	Min. : 9.71	Min. : 43.79	Min. :
143.5					
##	M:212	1st Qu.:11.700	1st Qu.:16.17	1st Qu.: 75.17	1st Qu.:
420.3					
##		Median :13.370	Median :18.84	Median : 86.24	Median :
551.1					
##		Mean :14.127	Mean :19.29	Mean : 91.97	Mean :
654.9					
##		3rd Qu.:15.780	3rd Qu.:21.80	3rd Qu.:104.10	3rd Qu.:
782.7					
##		Max. :28.110	Max. :39.28	Max. :188.50	Max. :
:2501.0					
##	smoothness_mean	compactness_mean	concavity_mean	concave.points_mean	
##	Min. :0.05263	Min. :0.01938	Min. :0.00000	Min. :0.00000	
##	1st Qu.:0.08637	1st Qu.:0.06492	1st Qu.:0.02956	1st Qu.:0.02031	
##	Median :0.09587	Median :0.09263	Median :0.06154	Median :0.03350	
##	Mean :0.09636	Mean :0.10434	Mean :0.08880	Mean :0.04892	
##	3rd Qu.:0.10530	3rd Qu.:0.13040	3rd Qu.:0.13070	3rd Qu.:0.07400	
##	Max. :0.16340	Max. :0.34540	Max. :0.42680	Max. :0.20120	
##	symmetry_mean	fractal_dimension_mean			
##	Min. :0.1060	Min. :0.04996			
##	1st Qu.:0.1619	1st Qu.:0.05770			
##	Median :0.1792	Median :0.06154			
##	Mean :0.1812	Mean :0.06280			
##	3rd Qu.:0.1957	3rd Qu.:0.06612			
##	Max. :0.3040	Max. :0.09744			

```
Cancer3Final$diagnosis<- as.numeric(Cancer3Final$diagnosis)
Cancer3Final$diagnosis[1:569] <- Cancer3Final$diagnosis[1:569] - 1
Cancer3Final$diagnosis
```

```
## [1] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1
1 1 1
## [38] 0 1 1 1 1 1 1 1 1 0 1 0 0 0 0 0 1 1 0 1 1 0 0 0 0 1 0 1 1 0 0 0 0 1
0 1 1
## [75] 0 1 0 1 1 0 0 0 1 1 0 1 1 1 0 0 0 1 0 0 1 1 0 0 0 1 1 0 0 0 0 1 0 0
1 0 0
## [112] 0 0 0 0 0 0 1 1 1 0 1 1 0 0 0 1 1 0 1 0 1 1 0 1 1 0 0 1 0 0 1 0 0 0
0 1 0
## [149] 0 0 0 0 0 0 0 0 1 0 0 0 0 1 1 0 1 0 0 1 1 0 0 1 1 0 0 0 0 1 0 0 1 1
1 0 1
## [186] 0 1 0 0 0 1 0 0 1 1 0 1 1 1 1 0 1 1 1 0 1 0 1 0 0 1 0 1 1 1 1 0 0 1
1 0 0
## [223] 0 1 0 0 0 0 0 1 1 0 0 1 0 0 1 1 0 1 0 0 0 0 1 0 0 0 0 0 1 0 1 1 1 1
```

```

1 1 1
## [260] 1 1 1 1 1 1 1 0 0 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 0 0 0 0 0 0
0 0 0
## [297] 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 1 0 1 0 0 0 0 1 1
1 0 0
## [334] 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0 0 1 1
0 1 1
## [371] 1 0 1 1 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 1 0 0 1 1 0 0 0 0 0 0 1 0 0 0
0 0 0
## [408] 0 1 0 0 0 0 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 1 0 1 0 0 0 0 0
1 0 0
## [445] 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0
0 1 0
## [482] 0 0 0 0 0 0 1 0 1 0 0 1 0 0 0 0 0 1 1 0 1 0 1 0 0 0 0 0 1 0 0 1 0 1
0 1 1
## [519] 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0
## [556] 0 0 0 0 0 0 0 1 1 1 1 1 1 1 0

```

##split the data set into training and eval

```

s = sample(399,170)
eval= Cancer3Final[s,] # eval with 60%
train= Cancer3Final[-s,] # train with 40%

```

regression on train set

```

res = glm(diagnosis~.,family=binomial,data=train)

```

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

```

summary(res)

```

```

##

```

```

## Call:

```

```

## glm(formula = diagnosis ~ ., family = binomial, data = train)

```

```

##

```

```

## Deviance Residuals:

```

```

##      Min       1Q   Median       3Q      Max
## -1.75729  -0.10560  -0.02162   0.00343   2.87650

```

```

##

```

```

## Coefficients:

```

```

##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -19.48319    20.40369  -0.955   0.3396
## radius_mean     2.00643     4.89863   0.410   0.6821
## texture_mean     0.41013     0.09039   4.537 0.0000057 ***
## perimeter_mean  -0.51230     0.66627  -0.769   0.4419
## area_mean       0.02731     0.02400   1.138   0.2552
## smoothness_mean 58.10482    43.87817   1.324   0.1854
## compactness_mean 8.17211    31.21603   0.262   0.7935
## concavity_mean   6.35997    10.95567   0.581   0.5616
## concave.points_mean 96.60006    42.36349   2.280   0.0226 *
## symmetry_mean   30.10462    14.05010   2.143   0.0321 *

```

```

## fractal_dimension_mean -80.03733 122.29429 -0.654 0.5128
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 521.841 on 398 degrees of freedom
## Residual deviance: 85.751 on 388 degrees of freedom
## AIC: 107.75
##
## Number of Fisher Scoring iterations: 9

drop1(res)

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## Single term deletions
##
## Model:
## diagnosis ~ radius_mean + texture_mean + perimeter_mean + area_mean +
## smoothness_mean + compactness_mean + concavity_mean +
## concave.points_mean +
## symmetry_mean + fractal_dimension_mean
##
## Df Deviance AIC
## <none> 85.751 107.75
## radius_mean 1 85.917 105.92
## texture_mean 1 116.902 136.90
## perimeter_mean 1 86.349 106.35
## area_mean 1 86.994 106.99
## smoothness_mean 1 87.579 107.58
## compactness_mean 1 85.820 105.82
## concavity_mean 1 86.091 106.09
## concave.points_mean 1 91.722 111.72

```



```

## symmetry_mean          1    90.561 110.56
## fractal_dimension_mean  1    86.185 106.19

reg1 = glm(diagnosis~.-compactness_mean, family = binomial, data=train)
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
drop1(reg1)
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Single term deletions
##
## Model:
## diagnosis ~ (radius_mean + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
##      concave.points_mean +
##      symmetry_mean + fractal_dimension_mean) - compactness_mean
##              Df Deviance    AIC
## <none>              85.820 105.82
## radius_mean          1    85.920 103.92
## texture_mean          1   117.146 135.15
## perimeter_mean        1    86.607 104.61
## area_mean             1    86.999 105.00
## smoothness_mean       1    87.685 105.69
## concavity_mean         1    86.133 104.13
## concave.points_mean    1    91.853 109.85
## symmetry_mean          1    91.040 109.04
## fractal_dimension_mean  1    86.248 104.25

reg2 = glm(diagnosis~.-radius_mean-compactness_mean, family = binomial,
data=train)
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
drop1(reg2)

```

```

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## Single term deletions
##
## Model:
## diagnosis ~ (radius_mean + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
##      concave.points_mean +
##      symmetry_mean + fractal_dimension_mean) - radius_mean -
## compactness_mean
##
##           Df Deviance    AIC
## <none>           85.920 103.92
## texture_mean      1  117.450 133.45
## perimeter_mean    1   87.636 103.64
## area_mean         1   90.346 106.35
## smoothness_mean   1   88.063 104.06
## concavity_mean     1   86.169 102.17
## concave.points_mean 1   91.865 107.86
## symmetry_mean      1   91.055 107.06
## fractal_dimension_mean 1   86.808 102.81

reg3 = glm(diagnosis~.-radius_mean-compactness_mean-concavity_mean, family =
binomial, data=train)

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

drop1(reg3)

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## Single term deletions
##
## Model:
## diagnosis ~ (radius_mean + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
##      concave.points_mean +
##      symmetry_mean + fractal_dimension_mean) - radius_mean -
##      compactness_mean -
##      concavity_mean
##
##          Df Deviance    AIC
## <none>          86.169 102.17
## texture_mean      1  117.773 131.77
## perimeter_mean    1   88.524 102.52
## area_mean         1   91.200 105.20
## smoothness_mean   1   88.430 102.43
## concave.points_mean 1  105.009 119.01
## symmetry_mean     1   92.404 106.40
## fractal_dimension_mean 1   86.809 100.81

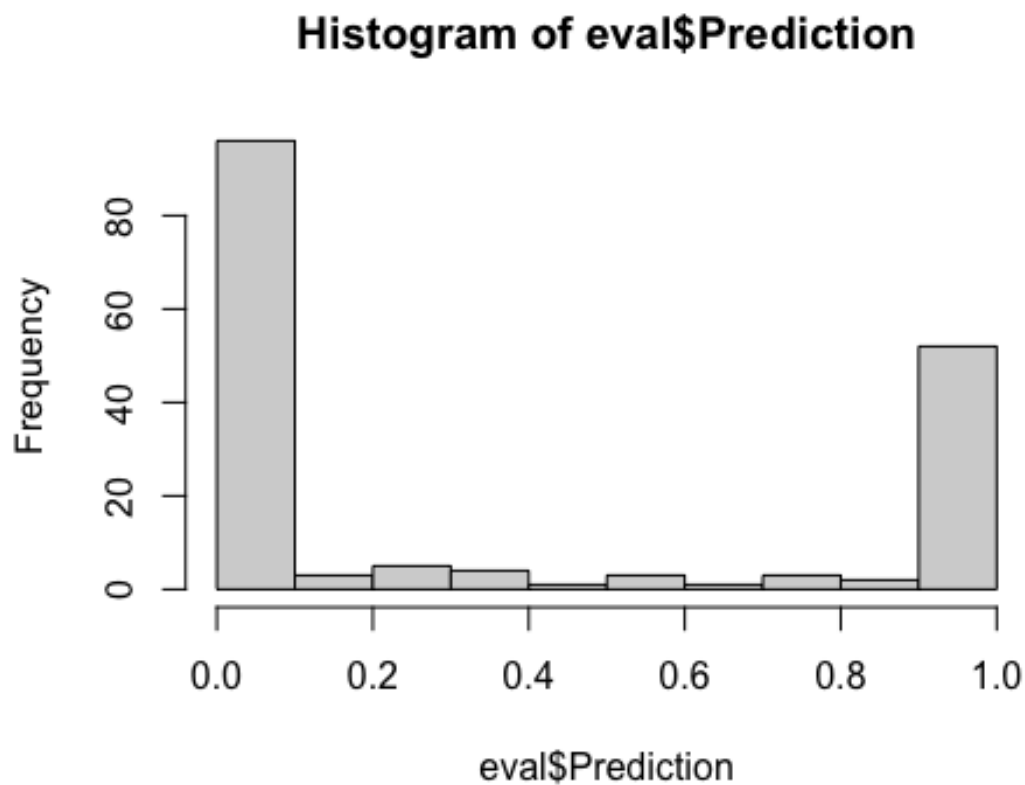
eval$Prediction=predict(reg3,eval,type="response")
eval[1:20,]

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
## smoothness_mean
## 324          1      20.34      21.51      135.90      1264.0
## 0.11700
## 167          0      10.80       9.71      68.77      357.6
## 0.09594
## 129          0      15.10      16.39      99.58      674.5
## 0.11500
## 299          0      14.26      18.17      91.22      633.1
## 0.06576
## 270          0      10.71      20.39      69.50      344.9
## 0.10820
## 187          1      18.31      18.58     118.60     1041.0
## 0.08588
## 307          0      13.20      15.82      84.07      537.3
## 0.08511
## 85           0      12.00      15.65      76.95      443.3
## 0.09723
## 277          0      11.33      14.16      71.79      396.6
## 0.09379
## 362          0      13.30      21.57      85.24      546.1
## 0.08582
## 330          1      16.26      21.88     107.50      826.8
## 0.11650
## 263          1      17.29      22.13     114.40      947.8
```

0.08999					
## 329	1	16.27	20.71	106.90	813.7
0.11690					
## 79	1	20.18	23.97	143.70	1245.0
0.12860					
## 213	1	28.11	18.47	188.50	2499.0
0.11420					
## 37	1	14.25	21.72	93.63	633.0
0.09823					
## 105	0	10.49	19.29	67.41	336.1
0.09989					
## 217	0	11.89	18.35	77.32	432.2
0.09363					
## 366	1	20.44	21.78	133.80	1293.0
0.09150					
## 165	1	23.27	22.04	152.10	1686.0
0.08439					
##	compactness_mean	concavity_mean	concave.points_mean	symmetry_mean	
## 324	0.18750	0.256500	0.150400	0.2569	
## 167	0.05736	0.025310	0.016980	0.1381	
## 129	0.18070	0.113800	0.085340	0.2001	
## 299	0.05220	0.024750	0.013740	0.1635	
## 270	0.12890	0.084480	0.028670	0.1668	
## 187	0.08468	0.081690	0.058140	0.1621	
## 307	0.05251	0.001461	0.003261	0.1632	
## 85	0.07165	0.041510	0.018630	0.2079	
## 277	0.03872	0.001487	0.003333	0.1954	
## 362	0.06373	0.033440	0.024240	0.1815	
## 330	0.12830	0.179900	0.079810	0.1869	
## 263	0.12730	0.096970	0.075070	0.2108	
## 329	0.13190	0.147800	0.084880	0.1948	
## 79	0.34540	0.375400	0.160400	0.2906	
## 213	0.15160	0.320100	0.159500	0.1648	
## 37	0.10980	0.131900	0.055980	0.1885	
## 105	0.08578	0.029950	0.012010	0.2217	
## 217	0.11540	0.066360	0.031420	0.1967	
## 366	0.11310	0.097990	0.077850	0.1618	
## 165	0.11450	0.132400	0.097020	0.1801	
##	fractal_dimension_mean	Prediction			
## 324	0.06670	0.99999999890			
## 167	0.06400	0.000005663968			
## 129	0.06467	0.935054132705			
## 299	0.05586	0.001388255956			
## 270	0.06862	0.002878512903			
## 187	0.05425	0.994464307003			
## 307	0.05894	0.000116760336			
## 85	0.05968	0.001498298902			
## 277	0.05821	0.000087465360			
## 362	0.05696	0.023578625180			
## 330	0.06532	0.998740498905			

```
## 263      0.05464 0.999762823565
## 329      0.06277 0.999002574889
## 79       0.08142 0.999999999893
## 213      0.05525 1.000000000000
## 37       0.06125 0.713860607256
## 105      0.06481 0.001905054820
## 217      0.06314 0.005573390666
## 366      0.05557 0.999997428743
## 165      0.05553 0.999999999916
```

```
hist(eval$Prediction)
```



```
plot(eval$diagnosis~eval$Prediction)
```



```
0 0 0
## [112] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0
## [149] 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0
```

Where was the model wrong?

```
eval[1:20,]
```

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean
## 324	1	20.34	21.51	135.90	1264.0	0.11700
## 167	0	10.80	9.71	68.77	357.6	0.09594
## 129	0	15.10	16.39	99.58	674.5	0.11500
## 299	0	14.26	18.17	91.22	633.1	0.06576
## 270	0	10.71	20.39	69.50	344.9	0.10820
## 187	1	18.31	18.58	118.60	1041.0	0.08588
## 307	0	13.20	15.82	84.07	537.3	0.08511
## 85	0	12.00	15.65	76.95	443.3	0.09723
## 277	0	11.33	14.16	71.79	396.6	0.09379
## 362	0	13.30	21.57	85.24	546.1	0.08582
## 330	1	16.26	21.88	107.50	826.8	0.11650
## 263	1	17.29	22.13	114.40	947.8	0.08999
## 329	1	16.27	20.71	106.90	813.7	0.11690
## 79	1	20.18	23.97	143.70	1245.0	0.12860
## 213	1	28.11	18.47	188.50	2499.0	0.11420
## 37	1	14.25	21.72	93.63	633.0	0.09823
## 105	0	10.49	19.29	67.41	336.1	0.09989
## 217	0	11.89	18.35	77.32	432.2	0.09363
## 366	1	20.44	21.78	133.80	1293.0	0.09150
## 165	1	23.27	22.04	152.10	1686.0	

```

0.08439
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 324      0.18750      0.256500      0.150400      0.2569
## 167      0.05736      0.025310      0.016980      0.1381
## 129      0.18070      0.113800      0.085340      0.2001
## 299      0.05220      0.024750      0.013740      0.1635
## 270      0.12890      0.084480      0.028670      0.1668
## 187      0.08468      0.081690      0.058140      0.1621
## 307      0.05251      0.001461      0.003261      0.1632
## 85       0.07165      0.041510      0.018630      0.2079
## 277      0.03872      0.001487      0.003333      0.1954
## 362      0.06373      0.033440      0.024240      0.1815
## 330      0.12830      0.179900      0.079810      0.1869
## 263      0.12730      0.096970      0.075070      0.2108
## 329      0.13190      0.147800      0.084880      0.1948
## 79       0.34540      0.375400      0.160400      0.2906
## 213      0.15160      0.320100      0.159500      0.1648
## 37       0.10980      0.131900      0.055980      0.1885
## 105      0.08578      0.029950      0.012010      0.2217
## 217      0.11540      0.066360      0.031420      0.1967
## 366      0.11310      0.097990      0.077850      0.1618
## 165      0.11450      0.132400      0.097020      0.1801
##      fractal_dimension_mean      Prediction falsePositive falseNegative
## 324      0.06670 0.999999999890      0      0
## 167      0.06400 0.000005663968      0      0
## 129      0.06467 0.935054132705      1      0
## 299      0.05586 0.001388255956      0      0
## 270      0.06862 0.002878512903      0      0
## 187      0.05425 0.994464307003      0      0
## 307      0.05894 0.000116760336      0      0
## 85       0.05968 0.001498298902      0      0
## 277      0.05821 0.000087465360      0      0
## 362      0.05696 0.023578625180      0      0
## 330      0.06532 0.998740498905      0      0
## 263      0.05464 0.999762823565      0      0
## 329      0.06277 0.999002574889      0      0
## 79       0.08142 0.999999999893      0      0
## 213      0.05525 1.000000000000      0      0
## 37       0.06125 0.713860607256      0      0
## 105      0.06481 0.001905054820      0      0
## 217      0.06314 0.005573390666      0      0
## 366      0.05557 0.999997428743      0      0
## 165      0.05553 0.999999999916      0      0

```

```

# Here we FILTER false positives and negatives from the data
falsePositive = eval[eval$falsePositive==1,]
falseNegative = eval[eval$falseNegative==1,]
falsePositive # This displays it on screen

```



```

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
smoothness_mean
## 129          0          15.10          16.39          99.58          674.5
0.1150
## 276          0          11.89          17.36          76.20          435.6
0.1225
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 129          0.1807          0.11380          0.08534          0.2001
## 276          0.0721          0.05929          0.07404          0.2015
##      fractal_dimension_mean Prediction falsePositive falseNegative
## 129          0.06467 0.9350541          1          0
## 276          0.05875 0.7974453          1          0

falseNegative

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
smoothness_mean
## 42          1          10.95          21.35          71.90          371.1
0.12270
## 40          1          13.48          20.82          88.40          559.2
0.10160
## 172          1          13.43          19.63          85.84          565.4
0.09048
## 298          1          11.76          18.14          75.00          431.1
0.09968
## 45          1          13.17          21.81          85.42          531.5
0.09714
## 206          1          15.12          16.68          98.78          716.6
0.08876
## 380          1          11.08          18.83          73.30          361.6
0.12160
## 127          1          13.61          24.69          87.76          572.6
0.09258
## 41          1          13.44          21.58          86.18          563.0
0.08162
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 42          0.12180          0.10440          0.05669          0.1895
## 40          0.12550          0.10630          0.05439          0.1720
## 172          0.06288          0.05858          0.03438          0.1598
## 298          0.05914          0.02685          0.03515          0.1619
## 45          0.10470          0.08259          0.05252          0.1746
## 206          0.09588          0.07550          0.04079          0.1594
## 380          0.21540          0.16890          0.06367          0.2196
## 127          0.07862          0.05285          0.03085          0.1761
## 41          0.06031          0.03110          0.02031          0.1784
##      fractal_dimension_mean Prediction falsePositive falseNegative
## 42          0.06870 0.320989574          0          1
## 40          0.06419 0.251576794          0          1
## 172          0.05671 0.033414894          0          1
## 298          0.06287 0.006573481          0          1

```

```

## 45          0.06177 0.277845095          0          1
## 206          0.05986 0.055791303          0          1
## 380          0.07950 0.202029928          0          1
## 127          0.06130 0.160224232          0          1
## 41           0.05587 0.016167469          0          1

error = (nrow(falseNegative) + nrow(falsePositive))/nrow(eval)
error

## [1] 0.06470588

## Our model got ~ 6.47% of the predictions wrong.

## PLOT THE LIFT CURVE
eval[1:3,]

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
## smoothness_mean
## 324          1          20.34          21.51          135.90      1264.0
## 0.11700
## 167          0          10.80           9.71           68.77       357.6
## 0.09594
## 129          0          15.10          16.39           99.58       674.5
## 0.11500
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 324          0.18750          0.25650          0.15040          0.2569
## 167          0.05736          0.02531          0.01698          0.1381
## 129          0.18070          0.11380          0.08534          0.2001
##      fractal_dimension_mean      Prediction falsePositive falseNegative
## 324          0.06670 0.999999999890          0          0
## 167          0.06400 0.000005663968          0          0
## 129          0.06467 0.935054132705          1          0

evalOrdered = eval[order(eval$Prediction, decreasing = TRUE),]
evalOrdered[1:3,]

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
## smoothness_mean
## 213          1          28.11          18.47          188.5       2499
## 0.1142
## 340          1          23.51          24.27          155.1       1747
## 0.1069
## 181          1          27.22          21.87          182.1       2250
## 0.1094
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 213          0.1516          0.3201          0.1595          0.1648
## 340          0.1283          0.2308          0.1410          0.1797
## 181          0.1914          0.2871          0.1878          0.1800
##      fractal_dimension_mean      Prediction falsePositive falseNegative
## 213          0.05525          1          0          0

```

```

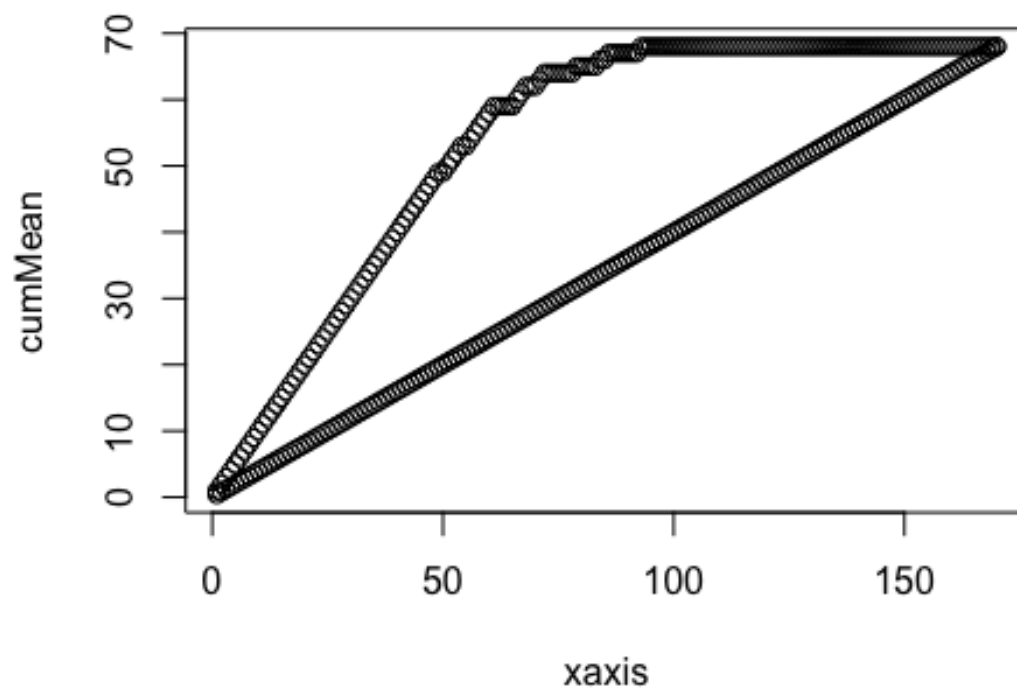
## 340          0.05506          1          0          0
## 181          0.05770          1          0          0

# dummy variables
xaxis = NULL
cumMean = NULL # cumulative mean
cumLift = NULL # cumulative lift
meanResponse = mean(eval$diagnosis) # mean curve
meanResponse # Our model predicts the results better than just guessing by
~37%

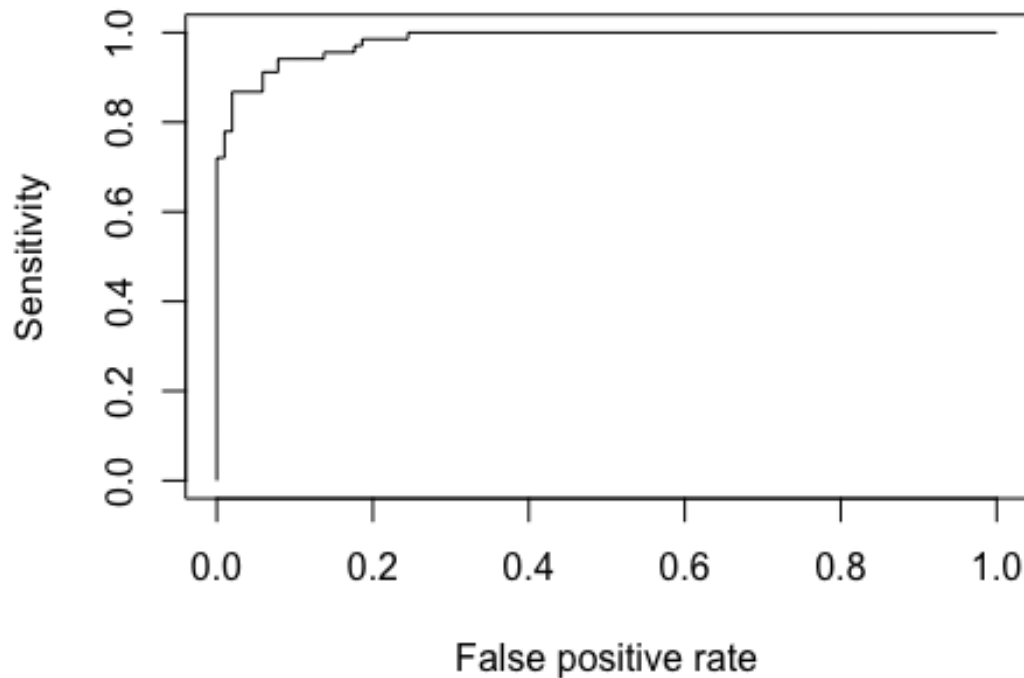
## [1] 0.4

# initiate variables
xaxis[1] = 1
cumMean[1] = meanResponse
cumLift[1] = evalOrdered$diagnosis[1] # We are most interested in the case
where we assume we are right
# repeat for all rows
for (i in 2:nrow(evalOrdered)){
  xaxis[i] = i
  cumMean[i] = cumMean[i-1]+meanResponse
  cumLift[i] = cumLift[i-1]+evalOrdered$diagnosis[i]
}
# At this point we need to check the cumulative lift in the values section.
In the beginning it goes up by 1, this is good sign.
plot(cumMean~xaxis) # plots the mean line
points(cumLift~xaxis) # plots the lift curve

```



```
## HERE WE PLOT THE ROC CURVE ##  
library(ROCR)  
prediction2 = prediction(eval$Prediction, eval$diagnosis)  
roc = performance(prediction2, "sens", "fpr")  
plot(roc) #this checks for different cutt off points
```

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

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[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
Cancer3Final$CV_prediction = CV
```

repeating previous steps

false positive, false negatives, misdiagnosisification rate

```
false_positive_cv=Cancer3Final[Cancer3Final$CV_prediction >= 0.5 &
Cancer3Final$diagnosis==0,]
false_positive_cv
```

```
##      diagnosis radius_mean texture_mean perimeter_mean area_mean
smoothness_mean
## 90          0      14.640      15.24      95.77      651.9
0.11320
## 113         0      14.260      19.65      97.83      629.9
0.07837
## 129         0      15.100      16.39      99.58      674.5
0.11500
## 153         0       9.731      15.34      63.78      300.2
0.10720
## 158         0      16.840      19.46     108.40      880.2
0.07445
## 239         0      14.220      27.85      92.55      623.9
0.08223
## 276         0      11.890      17.36      76.20      435.6
0.12250
## 364         0      16.500      18.29     106.60      838.1
0.09686
## 456         0      13.380      30.72      86.34      557.2
0.09245
## 492         0      17.850      13.23     114.60      992.1
```

```

0.07838
## 509      0      16.300      15.70      104.70      819.8
0.09427
## 529      0      13.940      13.17      90.31      594.2
0.12480
## 538      0      11.690      24.44      76.37      406.4
0.12360
## 542      0      14.470      24.99      95.81      656.4
0.08837
## 561      0      14.050      27.15      91.38      600.4
0.09929
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 90      0.13390      0.09966      0.07064      0.2116
## 113     0.22330      0.30030      0.07798      0.1704
## 129     0.18070      0.11380      0.08534      0.2001
## 153     0.15990      0.41080      0.07857      0.2548
## 158     0.07223      0.05150      0.02771      0.1844
## 239     0.10390      0.11030      0.04408      0.1342
## 276     0.07210      0.05929      0.07404      0.2015
## 364     0.08468      0.05862      0.04835      0.1495
## 456     0.07426      0.02819      0.03264      0.1375
## 492     0.06217      0.04445      0.04178      0.1220
## 509     0.06712      0.05526      0.04563      0.1711
## 529     0.09755      0.10100      0.06615      0.1976
## 538     0.15520      0.04515      0.04531      0.2131
## 542     0.12300      0.10090      0.03890      0.1872
## 561     0.11260      0.04462      0.04304      0.1537
##      fractal_dimension_mean CV_prediction
## 90      0.06346      0.7098976
## 113     0.07769      0.7472802
## 129     0.06467      0.9241708
## 153     0.09296      0.9860191
## 158     0.05268      0.6277209
## 239     0.06129      0.6720621
## 276     0.05875      0.8428632
## 364     0.05593      0.8149527
## 456     0.06016      0.7077005
## 492     0.05243      0.6833568
## 509     0.05657      0.5281348
## 529     0.06457      0.5213198
## 538     0.07405      0.5353958
## 542     0.06341      0.7446412
## 561     0.06171      0.6985200

nrow(false_positive_cv)

## [1] 15

```

```

false_negative_cv=Cancer3Final[Cancer3Final$CV_prediction < 0.5 &
Cancer3Final$diagnosis==1,]
false_negative_cv

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
smoothness_mean
## 32          1         11.84         18.70          77.93        440.6
0.11090
## 40          1         13.48         20.82          88.40        559.2
0.10160
## 41          1         13.44         21.58          86.18        563.0
0.08162
## 42          1         10.95         21.35          71.90        371.1
0.12270
## 44          1         13.28         20.28          87.32        545.2
0.10410
## 45          1         13.17         21.81          85.42        531.5
0.09714
## 55          1         15.10         22.02          97.26        712.8
0.09056
## 74          1         13.80         15.79          90.43        584.1
0.10070
## 127         1         13.61         24.69          87.76        572.6
0.09258
## 136         1         12.77         22.47          81.72        506.3
0.09055
## 147         1         11.80         16.58          78.99        432.0
0.10910
## 172         1         13.43         19.63          85.84        565.4
0.09048
## 185         1         15.28         22.41          98.92        710.6
0.09057
## 206         1         15.12         16.68          98.78        716.6
0.08876
## 216         1         13.86         16.93          90.96        578.9
0.10260
## 256         1         13.96         17.05          91.43        602.4
0.10960
## 264         1         15.61         19.38         100.00        758.6
0.07840
## 298         1         11.76         18.14          75.00        431.1
0.09968
## 380         1         11.08         18.83          73.30        361.6
0.12160
## 490         1         16.69         20.20         107.10        857.6
0.07497
## 515         1         15.05         19.07          97.26        701.9
0.09215
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 32          0.15160          0.12180          0.05182          0.2301

```

```
## 40      0.12550      0.10630      0.05439      0.1720
## 41      0.06031      0.03110      0.02031      0.1784
## 42      0.12180      0.10440      0.05669      0.1895
## 44      0.14360      0.09847      0.06158      0.1974
## 45      0.10470      0.08259      0.05252      0.1746
## 55      0.07081      0.05253      0.03334      0.1616
## 74      0.12800      0.07789      0.05069      0.1662
## 127     0.07862      0.05285      0.03085      0.1761
## 136     0.05761      0.04711      0.02704      0.1585
## 147     0.17000      0.16590      0.07415      0.2678
## 172     0.06288      0.05858      0.03438      0.1598
## 185     0.10520      0.05375      0.03263      0.1727
## 206     0.09588      0.07550      0.04079      0.1594
## 216     0.15170      0.09901      0.05602      0.2106
## 256     0.12790      0.09789      0.05246      0.1908
## 264     0.05616      0.04209      0.02847      0.1547
## 298     0.05914      0.02685      0.03515      0.1619
## 380     0.21540      0.16890      0.06367      0.2196
## 490     0.07112      0.03649      0.02307      0.1846
## 515     0.08597      0.07486      0.04335      0.1561
```

```
## fractal_dimension_mean CV_prediction
## 32      0.07799      0.161266765
## 40      0.06419      0.354805225
## 41      0.05587      0.022660394
## 42      0.06870      0.497252979
## 44      0.06782      0.468854531
## 45      0.06177      0.280729964
## 55      0.05684      0.433122073
## 74      0.06566      0.036142521
## 127     0.06130      0.226344915
## 136     0.06065      0.033237848
## 147     0.07371      0.426616845
## 172     0.05671      0.054728837
## 185     0.06317      0.230289830
## 206     0.05986      0.094737899
## 216     0.06916      0.109618230
## 256     0.06130      0.318913742
## 264     0.05443      0.109278719
## 298     0.06287      0.008123391
## 380     0.07950      0.134260079
## 490     0.05325      0.297966490
## 515     0.05915      0.261394798
```

```
nrow(false_negative_cv)
```

```
## [1] 21
```

```
error_cv = (nrow(false_negative_cv) +
nrow(false_positive_cv))/nrow(Cancer3Final)
error_cv
```



```
## [1] 0.06326889
```

lift curve

```
evalOrderedCV = Cancer3Final[order(Cancer3Final$CV_prediction, decreasing =  
TRUE),]  
evalOrderedCV
```

##	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean
## 83	1	25.220	24.91	171.50	1878.0	0.10630
## 123	1	24.250	20.20	166.20	1761.0	0.14470
## 181	1	27.220	21.87	182.10	2250.0	0.10940
## 203	1	23.290	26.67	158.90	1685.0	0.11410
## 213	1	28.110	18.47	188.50	2499.0	0.11420
## 340	1	23.510	24.27	155.10	1747.0	0.10690
## 353	1	25.730	17.46	174.20	2010.0	0.11490
## 462	1	27.420	26.27	186.90	2501.0	0.10840
## 522	1	24.630	21.60	165.50	1841.0	0.10300
## 109	1	22.270	19.67	152.80	1509.0	0.13260
## 237	1	23.210	26.97	153.50	1670.0	0.09509
## 565	1	21.560	22.39	142.00	1479.0	0.11100
## 370	1	22.010	21.90	147.20	1482.0	0.10630
## 394	1	21.610	22.28	144.40	1407.0	0.11670
## 568	1	20.600	29.33	140.10	1265.0	0.11780
## 504	1	23.090	19.83	152.10	1682.0	0.09342
## 165	1	23.270	22.04	152.10	1686.0	0.08439
## 564	1	20.920	25.09	143.00	1347.0	0.10990
## 79	1	20.180	23.97	143.70	1245.0	0.12860
## 266	1	20.730	31.12	135.70	1419.0	0.09469
## 251	1	20.940	23.56	138.90	1364.0	

0.10070					
## 182	1	21.090	26.57	142.70	1311.0
0.11410					
## 324	1	20.340	21.51	135.90	1264.0
0.11700					
## 273	1	21.750	20.99	147.30	1491.0
0.09401					
## 369	1	21.710	17.25	140.90	1546.0
0.09384					
## 536	1	20.550	20.86	137.80	1308.0
0.10460					
## 303	1	20.090	23.86	134.70	1247.0
0.10800					
## 84	1	19.100	26.29	129.10	1132.0
0.12150					
## 500	1	20.590	21.24	137.80	1320.0
0.10850					
## 450	1	21.100	20.52	138.10	1384.0
0.09684					
## 261	1	20.310	27.06	132.90	1288.0
0.10000					
## 24	1	21.160	23.04	137.20	1404.0
0.09428					
## 367	1	20.200	26.83	133.70	1234.0
0.09905					
## 566	1	20.130	28.25	131.20	1261.0
0.09780					
## 257	1	19.550	28.77	133.60	1207.0
0.09260					
## 3	1	19.690	21.25	130.00	1203.0
0.10960					
## 130	1	19.790	25.12	130.40	1192.0
0.10150					
## 163	1	19.590	18.15	130.70	1214.0
0.11200					
## 234	1	20.510	27.81	134.40	1319.0
0.09159					
## 433	1	20.180	19.54	133.80	1250.0
0.11330					
## 373	1	21.370	15.10	141.30	1386.0
0.10010					
## 31	1	18.630	25.11	124.80	1088.0
0.10640					
## 534	1	20.470	20.67	134.70	1299.0
0.09156					
## 19	1	19.810	22.15	130.00	1260.0
0.09831					
## 301	1	19.530	18.90	129.50	1217.0
0.11500					
## 220	1	19.530	32.47	128.00	1223.0

0.08420					
## 211	1	20.580	22.14	134.70	1290.0
0.09090					
## 344	1	19.680	21.68	129.90	1194.0
0.09797					
## 452	1	19.590	25.00	127.70	1191.0
0.10320					
## 281	1	19.160	26.60	126.20	1138.0
0.10200					
## 96	1	20.260	23.03	132.40	1264.0
0.09078					
## 390	1	19.550	23.21	128.90	1174.0
0.10100					
## 488	1	19.440	18.82	128.10	1167.0
0.10890					
## 240	1	17.460	39.28	113.40	920.6
0.09812					
## 374	1	20.640	17.35	134.80	1335.0
0.09446					
## 5	1	20.290	14.34	135.10	1297.0
0.10030					
## 219	1	19.800	21.56	129.70	1230.0
0.09383					
## 253	1	19.730	19.82	130.70	1206.0
0.10620					
## 245	1	19.400	23.50	129.10	1155.0
0.10270					
## 33	1	17.020	23.98	112.80	899.3
0.11970					
## 13	1	19.170	24.80	132.40	1123.0
0.09740					
## 401	1	17.910	21.02	124.40	994.0
0.12300					
## 366	1	20.440	21.78	133.80	1293.0
0.09150					
## 518	1	19.890	20.26	130.50	1214.0
0.10370					
## 34	1	19.270	26.47	127.90	1162.0
0.09401					
## 43	1	19.070	24.81	128.30	1104.0
0.09081					
## 322	1	20.160	19.66	131.10	1274.0
0.08020					
## 169	1	17.470	24.68	116.10	984.6
0.10490					
## 26	1	17.140	16.40	116.00	912.7
0.11860					
## 434	1	18.820	21.97	123.70	1110.0
0.10180					
## 2	1	20.570	17.77	132.90	1326.0

0.08474					
## 255	1	19.450	19.33	126.50	1169.0
0.10350					
## 238	1	20.480	21.46	132.50	1306.0
0.08355					
## 57	1	19.210	18.57	125.50	1152.0
0.10530					
## 54	1	18.220	18.70	120.30	1033.0
0.11480					
## 447	1	17.750	28.03	117.30	981.6
0.09997					
## 517	1	18.310	20.58	120.80	1052.0
0.10680					
## 283	1	19.400	18.18	127.20	1145.0
0.10370					
## 259	1	15.660	23.20	110.20	773.5
0.11090					
## 46	1	18.650	17.60	123.70	1076.0
0.10990					
## 1	1	17.990	10.38	122.80	1001.0
0.11840					
## 157	1	17.680	20.74	117.40	963.7
0.11150					
## 88	1	19.020	24.59	122.00	1076.0
0.09029					
## 71	1	18.940	21.31	123.60	1130.0
0.09009					
## 86	1	18.460	18.52	121.10	1075.0
0.09874					
## 260	1	15.530	33.56	103.70	744.9
0.10630					
## 162	1	19.190	15.94	126.30	1157.0
0.08694					
## 25	1	16.650	21.38	110.00	904.6
0.11210					
## 28	1	18.610	20.25	122.10	1094.0
0.09440					
## 199	1	19.180	22.49	127.50	1148.0
0.08523					
## 336	1	17.060	21.00	111.80	918.6
0.11190					
## 563	1	15.220	30.62	103.40	716.9
0.10480					
## 469	1	17.600	23.33	119.00	980.5
0.09289					
## 231	1	17.050	19.08	113.40	895.0
0.11410					
## 122	1	18.660	17.12	121.40	1077.0
0.10540					
## 73	1	17.200	24.52	114.20	929.4

0.10710					
## 135	1	18.450	21.91	120.20	1075.0
0.09430					
## 352	1	15.750	19.22	107.10	758.6
0.12430					
## 78	1	18.050	16.15	120.20	1006.0
0.10650					
## 499	1	18.490	17.52	121.30	1068.0
0.10120					
## 409	1	17.990	20.66	117.80	991.7
0.10360					
## 461	1	17.080	27.15	111.20	930.9
0.09898					
## 493	1	18.010	20.56	118.40	1007.0
0.10010					
## 128	1	19.000	18.91	123.40	1138.0
0.08217					
## 258	1	15.320	17.27	103.20	713.3
0.13350					
## 338	1	18.770	21.43	122.90	1092.0
0.09116					
## 119	1	15.780	22.91	105.70	782.6
0.11550					
## 214	1	17.420	25.56	114.50	948.0
0.10060					
## 7	1	18.250	19.98	119.60	1040.0
0.09463					
## 371	1	16.350	23.29	109.00	840.4
0.09742					
## 418	1	15.500	21.08	102.90	803.1
0.11200					
## 318	1	18.220	18.87	118.70	1027.0
0.09746					
## 263	1	17.290	22.13	114.40	947.8
0.08999					
## 278	1	18.810	19.98	120.90	1102.0
0.08923					
## 330	1	16.260	21.88	107.50	826.8
0.11650					
## 18	1	16.130	20.68	108.10	798.8
0.11700					
## 29	1	15.300	25.27	102.40	732.4
0.10820					
## 204	1	13.810	23.75	91.56	597.8
0.13230					
## 329	1	16.270	20.71	106.90	813.7
0.11690					
## 265	1	17.190	22.07	111.60	928.3
0.09726					
## 16	1	14.540	27.54	96.73	658.8

0.11390					
## 510	1	15.460	23.95	103.80	731.3
0.11830					
## 480	1	16.250	19.51	109.80	815.8
0.10260					
## 254	1	17.300	17.08	113.00	928.2
0.10080					
## 275	1	17.930	24.48	115.20	998.9
0.08855					
## 198	1	18.080	21.84	117.40	1024.0
0.07371					
## 393	1	15.490	19.97	102.40	744.7
0.11600					
## 442	1	17.270	25.42	112.40	928.8
0.08331					
## 66	1	14.780	23.94	97.40	668.3
0.11720					
## 431	1	14.900	22.53	102.10	685.0
0.09947					
## 178	1	16.460	20.11	109.30	832.9
0.09831					
## 567	1	16.600	28.08	108.30	858.1
0.08455					
## 120	1	17.950	20.01	114.20	982.0
0.08402					
## 187	1	18.310	18.58	118.60	1041.0
0.08588					
## 202	1	17.540	19.32	115.10	951.6
0.08968					
## 132	1	15.460	19.48	101.70	748.9
0.10920					
## 284	1	16.240	18.77	108.80	805.1
0.10660					
## 36	1	16.740	21.59	110.10	869.5
0.09610					
## 9	1	13.000	21.82	87.50	519.8
0.12730					
## 354	1	15.080	25.74	98.00	716.6
0.10240					
## 30	1	17.570	15.05	115.00	955.1
0.09847					
## 173	1	15.460	11.89	102.50	736.9
0.12570					
## 153	0	9.731	15.34	63.78	300.2
0.10720					
## 63	1	14.250	22.15	96.42	645.7
0.10490					
## 208	1	17.010	20.26	109.70	904.3
0.08772					
## 95	1	15.060	19.83	100.30	705.6

0.10390					
## 133	1	16.160	21.54	106.20	809.8
0.10080					
## 4	1	11.420	20.38	77.58	386.1
0.14250					
## 168	1	16.780	18.80	109.30	886.3
0.08865					
## 195	1	14.860	23.21	100.40	671.4
0.10440					
## 502	1	13.820	24.49	92.33	595.9
0.11620					
## 58	1	14.710	21.59	95.55	656.9
0.11370					
## 35	1	16.130	17.88	107.00	807.2
0.10400					
## 445	1	18.030	16.85	117.50	990.0
0.08947					
## 76	1	16.070	19.65	104.10	817.7
0.09168					
## 118	1	14.870	16.67	98.64	682.5
0.11620					
## 139	1	14.950	17.57	96.85	678.1
0.11670					
## 27	1	14.580	21.53	97.41	644.8
0.10540					
## 10	1	12.460	24.04	83.97	475.9
0.11860					
## 92	1	15.370	22.76	100.20	728.2
0.09200					
## 14	1	15.850	23.95	103.70	782.7
0.08401					
## 15	1	13.730	22.61	93.60	578.3
0.11310					
## 106	1	13.110	15.56	87.21	530.2
0.13980					
## 224	1	15.750	20.25	102.60	761.3
0.10250					
## 23	1	15.340	14.26	102.50	704.4
0.10730					
## 197	1	13.770	22.29	90.63	588.9
0.12000					
## 415	1	15.130	29.81	96.71	719.5
0.08320					
## 262	1	17.350	23.06	111.00	933.1
0.08662					
## 537	1	14.270	22.55	93.77	629.8
0.10380					
## 129	0	15.100	16.39	99.58	674.5
0.11500					
## 191	1	14.220	23.12	94.37	609.9

0.10750					
## 65	1	12.680	23.84	82.69	499.0
0.11220					
## 513	1	13.400	20.52	88.64	556.7
0.11060					
## 142	1	16.110	18.05	105.10	813.0
0.09721					
## 12	1	15.780	17.89	103.60	781.0
0.09710					
## 215	1	14.190	23.81	92.87	610.7
0.09463					
## 276	0	11.890	17.36	76.20	435.6
0.12250					
## 183	1	15.700	20.31	101.20	766.6
0.09597					
## 364	0	16.500	18.29	106.60	838.1
0.09686					
## 230	1	12.830	22.33	85.26	503.2
0.10880					
## 48	1	13.170	18.66	85.98	534.6
0.11580					
## 37	1	14.250	21.72	93.63	633.0
0.09823					
## 331	1	16.030	15.51	105.80	793.2
0.09491					
## 8	1	13.710	20.83	90.20	577.9
0.11890					
## 113	0	14.260	19.65	97.83	629.9
0.07837					
## 87	1	14.480	21.46	94.25	648.2
0.09444					
## 542	0	14.470	24.99	95.81	656.4
0.08837					
## 386	1	14.600	23.29	93.97	664.7
0.08682					
## 194	1	12.340	26.86	81.15	477.4
0.10340					
## 90	0	14.640	15.24	95.77	651.9
0.11320					
## 456	0	13.380	30.72	86.34	557.2
0.09245					
## 561	0	14.050	27.15	91.38	600.4
0.09929					
## 17	1	14.680	20.13	94.74	684.5
0.09867					
## 492	0	17.850	13.23	114.60	992.1
0.07838					
## 239	0	14.220	27.85	92.55	623.9
0.08223					
## 200	1	14.450	20.22	94.49	642.7

0.09872					
## 101	1	13.610	24.98	88.05	582.7
0.09488					
## 6	1	12.450	15.70	82.57	477.1
0.12780					
## 158	0	16.840	19.46	108.40	880.2
0.07445					
## 39	1	14.990	25.20	95.54	698.8
0.09387					
## 11	1	16.020	23.24	102.70	797.8
0.08206					
## 436	1	13.980	19.62	91.12	599.5
0.10600					
## 538	0	11.690	24.44	76.37	406.4
0.12360					
## 509	0	16.300	15.70	104.70	819.8
0.09427					
## 100	1	14.420	19.77	94.48	642.5
0.09752					
## 529	0	13.940	13.17	90.31	594.2
0.12480					
## 42	1	10.950	21.35	71.90	371.1
0.12270					
## 376	0	16.170	16.07	106.30	788.5
0.09880					
## 291	0	14.410	19.73	96.03	651.0
0.08757					
## 44	1	13.280	20.28	87.32	545.2
0.10410					
## 543	0	14.740	25.42	94.70	668.6
0.08275					
## 496	0	14.870	20.21	96.12	680.9
0.09587					
## 414	0	14.990	22.11	97.53	693.7
0.08515					
## 55	1	15.100	22.02	97.26	712.8
0.09056					
## 147	1	11.800	16.58	78.99	432.0
0.10910					
## 151	0	13.000	20.78	83.51	519.4
0.11350					
## 134	0	15.710	13.93	102.00	761.7
0.09462					
## 544	0	13.210	28.06	84.88	538.4
0.08671					
## 40	1	13.480	20.82	88.40	559.2
0.10160					
## 77	0	13.530	10.94	87.91	559.2
0.12910					
## 292	0	14.960	19.10	97.03	687.3

0.08992					
## 91	0	14.620	24.02	94.57	662.7
0.08974					
## 256	1	13.960	17.05	91.43	602.4
0.10960					
## 397	0	13.510	18.89	88.10	558.1
0.10590					
## 407	0	16.140	14.86	104.30	800.0
0.09495					
## 490	1	16.690	20.20	107.10	857.6
0.07497					
## 446	0	11.990	24.89	77.61	441.3
0.10300					
## 45	1	13.170	21.81	85.42	531.5
0.09714					
## 457	0	11.630	29.29	74.87	415.1
0.09357					
## 209	0	13.110	22.54	87.02	529.4
0.10020					
## 515	1	15.050	19.07	97.26	701.9
0.09215					
## 454	0	14.530	13.98	93.86	644.2
0.10990					
## 357	0	13.050	18.59	85.09	512.0
0.10820					
## 555	0	12.880	28.92	82.50	514.3
0.08123					
## 501	0	15.040	16.74	98.73	689.4
0.09883					
## 185	1	15.280	22.41	98.92	710.6
0.09057					
## 472	0	12.040	28.14	76.85	449.9
0.08752					
## 485	0	15.730	11.28	102.80	747.2
0.10430					
## 127	1	13.610	24.69	87.76	572.6
0.09258					
## 559	0	14.590	22.68	96.39	657.1
0.08473					
## 233	0	11.220	33.81	70.79	386.8
0.07780					
## 422	0	14.690	13.98	98.22	656.1
0.10310					
## 463	0	14.400	26.99	92.25	646.1
0.06995					
## 289	0	11.260	19.96	73.72	394.1
0.08020					
## 112	0	12.630	20.76	82.15	480.4
0.09933					
## 505	0	9.268	12.87	61.49	248.7

0.16340					
## 32	1	11.840	18.70	77.93	440.6
0.11090					
## 69	0	9.029	17.33	58.79	250.5
0.10660					
## 553	0	12.770	29.43	81.35	507.9
0.08276					
## 470	0	11.620	18.18	76.38	408.8
0.11750					
## 149	0	14.440	15.18	93.97	640.1
0.09970					
## 82	0	13.340	15.86	86.49	520.0
0.10780					
## 380	1	11.080	18.83	73.30	361.6
0.12160					
## 170	0	14.970	16.95	96.22	685.9
0.09855					
## 424	0	13.660	19.13	89.46	575.3
0.09057					
## 226	0	14.340	13.47	92.51	641.2
0.09906					
## 50	0	13.490	22.30	86.91	561.0
0.08752					
## 244	0	13.750	23.77	88.54	590.0
0.08043					
## 216	1	13.860	16.93	90.96	578.9
0.10260					
## 264	1	15.610	19.38	100.00	758.6
0.07840					
## 556	0	10.290	27.61	65.67	321.4
0.09030					
## 546	0	13.620	23.23	87.19	573.2
0.09246					
## 458	0	13.210	25.25	84.10	537.9
0.08791					
## 341	0	14.420	16.54	94.15	641.2
0.09751					
## 519	0	12.880	18.22	84.45	493.1
0.12180					
## 206	1	15.120	16.68	98.78	716.6
0.08876					
## 161	0	11.750	20.18	76.10	419.8
0.10890					
## 377	0	10.570	20.22	70.15	338.3
0.09073					
## 453	0	12.000	28.23	76.77	442.5
0.08437					
## 514	0	14.580	13.66	94.29	658.8
0.09832					
## 124	0	14.500	10.89	94.28	640.7

0.11010					
## 560	0	11.510	23.93	74.52	403.5
0.09261					
## 449	0	14.530	19.34	94.25	659.7
0.08388					
## 148	0	14.950	18.77	97.84	689.5
0.08138					
## 201	0	12.230	19.56	78.54	461.0
0.09586					
## 497	0	12.650	18.17	82.69	485.6
0.10760					
## 172	1	13.430	19.63	85.84	565.4
0.09048					
## 205	0	12.470	18.60	81.09	481.9
0.09965					
## 467	0	13.140	20.74	85.98	536.9
0.08675					
## 459	0	13.000	25.13	82.61	520.2
0.08369					
## 378	0	13.460	28.21	85.89	562.1
0.07517					
## 319	0	9.042	18.90	60.07	244.5
0.09968					
## 527	0	13.460	18.75	87.44	551.1
0.10750					
## 20	0	13.540	14.36	87.46	566.3
0.09779					
## 477	0	14.200	20.53	92.41	618.4
0.08931					
## 381	0	11.270	12.96	73.16	386.3
0.12370					
## 448	0	14.800	17.66	95.88	674.8
0.09179					
## 166	0	14.970	19.76	95.50	690.2
0.08421					
## 89	0	12.360	21.80	79.78	466.1
0.08772					
## 107	0	11.640	18.33	75.17	412.5
0.11420					
## 249	0	10.650	25.22	68.01	347.0
0.09657					
## 507	0	12.220	20.04	79.47	453.1
0.10960					
## 362	0	13.300	21.57	85.24	546.1
0.08582					
## 74	1	13.800	15.79	90.43	584.1
0.10070					
## 62	0	8.598	20.98	54.66	221.8
0.12430					
## 94	0	13.450	18.30	86.60	555.1

0.10220					
## 524	0	13.710	18.68	88.73	571.0
0.09916					
## 136	1	12.770	22.47	81.72	506.3
0.09055					
## 222	0	13.560	13.90	88.59	561.3
0.10510					
## 280	0	13.850	15.18	88.99	587.4
0.09516					
## 474	0	12.270	29.97	77.42	465.4
0.07699					
## 164	0	12.340	22.22	79.85	464.5
0.10120					
## 228	0	15.000	15.51	97.45	684.5
0.08371					
## 486	0	12.450	16.41	82.85	476.7
0.09514					
## 246	0	10.480	19.86	66.72	337.7
0.10700					
## 484	0	13.700	17.64	87.76	571.1
0.09950					
## 503	0	12.540	16.32	81.25	476.3
0.11580					
## 520	0	12.750	16.70	82.51	493.8
0.11250					
## 423	0	11.610	16.02	75.46	408.2
0.10880					
## 236	0	14.030	21.25	89.79	603.4
0.09070					
## 552	0	11.130	22.44	71.49	378.4
0.09566					
## 487	0	14.640	16.85	94.21	666.0
0.08641					
## 356	0	12.560	19.07	81.92	485.8
0.08760					
## 483	0	13.470	14.06	87.32	546.3
0.10710					
## 81	0	11.450	20.97	73.81	401.5
0.11020					
## 435	0	14.860	16.94	94.89	673.7
0.08924					
## 192	0	12.770	21.41	82.02	507.4
0.08749					
## 241	0	13.640	15.60	87.38	575.3
0.09423					
## 531	0	11.750	17.56	75.89	422.9
0.10730					
## 416	0	11.890	21.17	76.39	433.8
0.09773					
## 41	1	13.440	21.58	86.18	563.0

0.08162					
## 540	0	7.691	25.44	48.34	170.4
0.08668					
## 545	0	13.870	20.70	89.77	584.8
0.09578					
## 395	0	12.100	17.72	78.07	446.2
0.10290					
## 460	0	9.755	28.20	61.68	290.9
0.07984					
## 104	0	9.876	19.40	63.95	298.3
0.10050					
## 114	0	10.510	20.19	68.64	334.2
0.11220					
## 229	0	12.620	23.97	81.35	496.4
0.07903					
## 270	0	10.710	20.39	69.50	344.9
0.10820					
## 332	0	12.980	19.35	84.52	514.0
0.09579					
## 345	0	11.710	15.45	75.03	420.3
0.11500					
## 508	0	11.060	17.12	71.25	366.5
0.11940					
## 425	0	9.742	19.12	61.93	289.7
0.10750					
## 506	0	9.676	13.14	64.12	272.5
0.12550					
## 417	0	9.405	21.70	59.60	271.2
0.10440					
## 116	0	11.930	21.53	76.53	438.6
0.09768					
## 300	0	10.510	23.09	66.85	334.2
0.10150					
## 439	0	13.850	19.60	88.68	592.6
0.08684					
## 489	0	11.680	16.17	75.49	420.5
0.11280					
## 140	0	11.280	13.39	73.00	384.8
0.11640					
## 155	0	13.150	15.34	85.31	538.9
0.09384					
## 516	0	11.340	18.61	72.76	391.2
0.10490					
## 473	0	14.920	14.93	96.45	686.9
0.08098					
## 466	0	13.240	20.13	86.87	542.9
0.08284					
## 348	0	14.760	14.74	94.87	668.7
0.08875					
## 512	0	14.810	14.70	94.66	680.7

0.08472					
## 268	0	13.590	21.84	87.16	561.0
0.07956					
## 532	0	11.670	20.02	75.21	416.2
0.10160					
## 152	0	8.219	20.70	53.27	203.9
0.09405					
## 287	0	11.940	20.76	77.87	441.0
0.08605					
## 482	0	13.900	19.24	88.73	602.9
0.07991					
## 97	0	12.180	17.84	77.79	451.1
0.10450					
## 217	0	11.890	18.35	77.32	432.2
0.09363					
## 293	0	12.950	16.02	83.14	513.7
0.10050					
## 21	0	13.080	15.71	85.63	520.0
0.10750					
## 396	0	14.060	17.18	89.75	609.1
0.08045					
## 558	0	9.423	27.88	59.26	271.3
0.08123					
## 428	0	10.800	21.98	68.79	359.9
0.08801					
## 521	0	9.295	13.90	59.96	257.8
0.13710					
## 80	0	12.860	18.00	83.19	506.3
0.09934					
## 398	0	12.800	17.46	83.05	508.3
0.08044					
## 306	0	11.600	24.49	74.23	417.2
0.07474					
## 210	0	15.270	12.91	98.17	725.5
0.08182					
## 70	0	12.780	16.49	81.37	502.5
0.09831					
## 302	0	12.460	19.89	80.43	471.3
0.08451					
## 56	0	11.520	18.75	73.34	409.0
0.09524					
## 38	0	13.030	18.42	82.61	523.8
0.08983					
## 438	0	14.040	15.98	89.78	611.2
0.08458					
## 432	0	12.400	17.68	81.47	467.8
0.10540					
## 410	0	12.270	17.92	78.41	466.1
0.08685					
## 267	0	10.600	18.95	69.28	346.4

0.09688					
## 437	0	12.870	19.54	82.67	509.2
0.09136					
## 232	0	11.320	27.08	71.76	395.7
0.06883					
## 384	0	12.390	17.48	80.64	462.9
0.10420					
## 298	1	11.760	18.14	75.00	431.1
0.09968					
## 243	0	11.300	18.19	73.93	389.4
0.09592					
## 326	0	12.670	17.30	81.25	489.9
0.10280					
## 333	0	11.220	19.86	71.94	387.3
0.10540					
## 321	0	10.250	16.18	66.52	324.2
0.10610					
## 372	0	15.190	13.21	97.65	711.8
0.07963					
## 420	0	11.160	21.41	70.95	380.3
0.10180					
## 363	0	12.760	18.84	81.87	496.6
0.09676					
## 550	0	10.820	24.21	68.89	361.6
0.08192					
## 406	0	10.940	18.59	70.39	370.0
0.10040					
## 562	0	11.200	29.37	70.67	386.0
0.07449					
## 408	0	12.850	21.37	82.63	514.5
0.07551					
## 67	0	9.465	21.01	60.11	269.4
0.10440					
## 105	0	10.490	19.29	67.41	336.1
0.09989					
## 269	0	12.870	16.21	82.38	512.2
0.09425					
## 350	0	11.950	14.96	77.23	426.7
0.11580					
## 207	0	9.876	17.27	62.92	295.4
0.10890					
## 188	0	11.710	17.19	74.68	420.3
0.09774					
## 323	0	12.860	13.32	82.82	504.8
0.11340					
## 312	0	14.610	15.69	92.68	664.9
0.07618					
## 412	0	11.040	16.83	70.92	373.2
0.10770					
## 491	0	12.250	22.44	78.18	466.5

0.08192					
## 252	0	11.500	18.45	73.28	407.4
0.09345					
## 476	0	12.830	15.73	82.89	506.9
0.09040					
## 103	0	12.180	20.52	77.22	458.7
0.08013					
## 177	0	9.904	18.06	64.60	302.4
0.09699					
## 554	0	9.333	21.94	59.01	264.0
0.09240					
## 171	0	12.320	12.39	78.85	464.1
0.10280					
## 137	0	11.710	16.67	74.72	423.6
0.10510					
## 290	0	11.370	18.89	72.17	396.0
0.08713					
## 49	0	12.050	14.63	78.04	449.3
0.10310					
## 368	0	12.210	18.02	78.31	458.4
0.09231					
## 225	0	13.270	17.02	84.55	546.4
0.08445					
## 465	0	13.170	18.22	84.28	537.3
0.07466					
## 61	0	10.170	14.88	64.55	311.9
0.11340					
## 375	0	13.690	16.07	87.84	579.1
0.08302					
## 189	0	11.810	17.39	75.27	428.9
0.10070					
## 156	0	12.250	17.94	78.27	460.3
0.08654					
## 248	0	12.890	14.11	84.95	512.2
0.08760					
## 539	0	7.729	25.49	47.98	178.8
0.08098					
## 404	0	12.940	16.17	83.18	507.6
0.09879					
## 250	0	11.520	14.93	73.87	406.3
0.10130					
## 85	0	12.000	15.65	76.95	443.3
0.09723					
## 110	0	11.340	21.26	72.48	396.5
0.08759					
## 455	0	12.620	17.15	80.62	492.9
0.08583					
## 440	0	14.020	15.66	89.59	606.5
0.07966					
## 379	0	13.660	15.15	88.27	580.6

0.08268					
## 51	0	11.760	21.60	74.72	427.9
0.08637					
## 146	0	11.900	14.65	78.11	432.8
0.11520					
## 533	0	13.680	16.33	87.76	575.5
0.09277					
## 144	0	12.900	15.92	83.74	512.2
0.08677					
## 471	0	9.667	18.49	61.49	289.1
0.08946					
## 75	0	12.310	16.52	79.19	470.9
0.09172					
## 339	0	10.050	17.53	64.41	310.8
0.10070					
## 495	0	13.160	20.54	84.06	538.7
0.07335					
## 535	0	10.960	17.62	70.79	365.6
0.09687					
## 143	0	11.430	17.31	73.66	398.0
0.10920					
## 441	0	10.970	17.20	71.73	371.5
0.08915					
## 304	0	10.490	18.61	66.86	334.3
0.10680					
## 358	0	13.870	16.21	88.52	593.7
0.08743					
## 285	0	12.890	15.70	84.08	516.6
0.07818					
## 327	0	14.110	12.88	90.03	616.5
0.09309					
## 223	0	10.180	17.53	65.12	313.1
0.10610					
## 196	0	12.910	16.33	82.53	516.4
0.07941					
## 403	0	12.960	18.29	84.18	525.2
0.07351					
## 138	0	11.430	15.39	73.06	399.8
0.09639					
## 343	0	11.060	14.96	71.49	373.9
0.10330					
## 421	0	11.570	19.04	74.20	409.7
0.08546					
## 294	0	11.850	17.46	75.54	432.7
0.08372					
## 383	0	12.050	22.72	78.75	447.8
0.06935					
## 126	0	13.850	17.21	88.44	588.7
0.08785					
## 131	0	12.190	13.29	79.08	455.8

0.10660					
## 557	0	10.160	19.59	64.73	311.7
0.10030					
## 279	0	13.590	17.84	86.24	572.3
0.07948					
## 498	0	12.470	17.31	80.45	480.1
0.08928					
## 150	0	13.740	17.91	88.12	585.0
0.07944					
## 299	0	14.260	18.17	91.22	633.1
0.06576					
## 59	0	13.050	19.31	82.61	527.2
0.08060					
## 400	0	11.800	17.26	75.26	431.9
0.09087					
## 68	0	11.310	19.04	71.80	394.1
0.08139					
## 311	0	11.700	19.11	74.33	418.7
0.08814					
## 530	0	12.070	13.44	77.83	445.2
0.11000					
## 115	0	8.726	15.83	55.84	230.9
0.11500					
## 444	0	10.570	18.32	66.82	340.9
0.08142					
## 411	0	11.360	17.57	72.49	399.8
0.08858					
## 315	0	8.597	18.60	54.09	221.2
0.10740					
## 108	0	12.360	18.54	79.01	466.7
0.08477					
## 360	0	9.436	18.32	59.82	278.6
0.10090					
## 125	0	13.370	16.39	86.10	553.5
0.07115					
## 479	0	11.490	14.59	73.99	404.9
0.10460					
## 385	0	13.280	13.72	85.79	541.8
0.08363					
## 365	0	13.400	16.95	85.48	552.4
0.07937					
## 212	0	11.840	18.94	75.51	428.0
0.08871					
## 478	0	13.900	16.62	88.97	599.4
0.06828					
## 430	0	12.720	17.67	80.98	501.3
0.07896					
## 464	0	11.600	18.36	73.88	412.7
0.08508					
## 53	0	11.940	18.24	75.71	437.6

0.08261					
## 481	0	12.160	18.03	78.29	455.3
0.09087					
## 98	0	9.787	19.94	62.11	294.5
0.10240					
## 541	0	11.540	14.44	74.65	402.9
0.09984					
## 111	0	9.777	16.99	62.50	290.2
0.10370					
## 296	0	13.770	13.27	88.06	582.7
0.09198					
## 221	0	13.650	13.16	87.88	568.9
0.09646					
## 93	0	13.270	14.76	84.74	551.7
0.07355					
## 342	0	9.606	16.84	61.64	280.5
0.08481					
## 451	0	11.870	21.54	76.83	432.0
0.06613					
## 179	0	13.010	22.22	82.01	526.4
0.06251					
## 443	0	13.780	15.79	88.37	585.9
0.08817					
## 427	0	10.480	14.98	67.49	333.6
0.09816					
## 335	0	12.300	19.02	77.88	464.4
0.08313					
## 349	0	11.470	16.03	73.02	402.7
0.09076					
## 286	0	12.580	18.40	79.83	489.0
0.08393					
## 475	0	10.880	15.62	70.41	358.9
0.10070					
## 247	0	13.200	17.43	84.13	541.6
0.07215					
## 405	0	12.340	14.95	78.29	469.1
0.08682					
## 174	0	11.080	14.71	70.21	372.7
0.10060					
## 272	0	11.290	13.04	72.23	388.0
0.09834					
## 305	0	11.460	18.16	73.59	403.1
0.08853					
## 313	0	12.760	13.37	82.29	504.1
0.08794					
## 282	0	11.740	14.02	74.24	427.3
0.07813					
## 227	0	10.440	15.46	66.62	329.6
0.10530					
## 548	0	10.260	16.58	65.85	320.8

0.08877					
## 190	0	12.300	15.90	78.83	463.7
0.08080					
## 549	0	9.683	19.34	61.05	285.7
0.08491					
## 392	0	8.734	16.84	55.27	234.3
0.10390					
## 274	0	9.742	15.67	61.50	289.9
0.09037					
## 347	0	12.060	18.90	76.66	445.3
0.08386					
## 389	0	11.270	15.50	73.38	392.0
0.08365					
## 359	0	8.878	15.49	56.74	241.0
0.08293					
## 295	0	12.720	13.78	81.78	492.1
0.09667					
## 361	0	12.540	18.07	79.42	491.9
0.07436					
## 388	0	13.880	16.16	88.37	596.6
0.07026					
## 426	0	10.030	21.28	63.19	307.3
0.08117					
## 218	0	10.200	17.48	65.05	321.2
0.08054					
## 523	0	11.260	19.83	71.30	388.1
0.08511					
## 52	0	13.640	16.34	87.21	571.8
0.07685					
## 102	0	6.981	13.43	43.79	143.5
0.11700					
## 271	0	14.290	16.82	90.30	632.6
0.06429					
## 528	0	12.340	12.27	78.94	468.5
0.09003					
## 320	0	12.430	17.00	78.60	477.3
0.07557					
## 387	0	12.210	14.09	78.78	462.0
0.08108					
## 117	0	8.950	15.76	58.74	245.2
0.09462					
## 154	0	11.150	13.08	70.87	381.9
0.09754					
## 337	0	12.990	14.23	84.08	514.3
0.09462					
## 569	0	7.760	24.54	47.92	181.0
0.05263					
## 547	0	10.320	16.35	65.31	324.9
0.09434					
## 325	0	12.200	15.21	78.01	457.9

0.08673					
## 72	0	8.888	14.64	58.79	244.0
0.09783					
## 99	0	11.600	12.84	74.34	412.6
0.08983					
## 551	0	10.860	21.48	68.51	360.5
0.07431					
## 307	0	13.200	15.82	84.07	537.3
0.08511					
## 184	0	11.410	14.92	73.53	402.0
0.09059					
## 346	0	10.260	14.71	66.20	321.6
0.09882					
## 413	0	9.397	21.68	59.75	268.8
0.07969					
## 159	0	12.060	12.74	76.84	448.6
0.09311					
## 382	0	11.040	14.93	70.67	372.7
0.07987					
## 351	0	11.660	17.07	73.70	421.0
0.07561					
## 525	0	9.847	15.68	63.00	293.2
0.09492					
## 22	0	9.504	12.44	60.34	273.9
0.10240					
## 64	0	9.173	13.86	59.20	260.9
0.07721					
## 47	0	8.196	16.84	51.71	201.9
0.08600					
## 277	0	11.330	14.16	71.79	396.6
0.09379					
## 316	0	12.490	16.85	79.19	481.6
0.08511					
## 310	0	13.050	13.84	82.71	530.6
0.08352					
## 468	0	9.668	18.10	61.06	286.3
0.08311					
## 419	0	12.700	12.17	80.88	495.0
0.08785					
## 511	0	11.740	14.69	76.31	426.0
0.08099					
## 526	0	8.571	13.10	54.53	221.3
0.10360					
## 121	0	11.410	10.82	73.34	403.3
0.09373					
## 391	0	10.260	12.22	65.75	321.6
0.09996					
## 429	0	11.130	16.62	70.47	381.1
0.08151					
## 235	0	9.567	15.91	60.21	279.6

0.08464					
## 242	0	12.420	15.04	78.61	476.5
0.07926					
## 186	0	10.080	15.11	63.76	317.5
0.09267					
## 180	0	12.810	13.06	81.29	508.8
0.08739					
## 402	0	11.930	10.91	76.14	442.7
0.08872					
## 328	0	12.030	17.93	76.09	446.0
0.07683					
## 175	0	10.660	15.15	67.49	349.6
0.08792					
## 317	0	12.180	14.08	77.25	461.4
0.07734					
## 334	0	11.250	14.78	71.38	390.0
0.08306					
## 355	0	11.140	14.07	71.24	384.6
0.07274					
## 309	0	13.500	12.71	85.69	566.2
0.07376					
## 176	0	8.671	14.45	54.42	227.2
0.09138					
## 145	0	10.750	14.97	68.26	355.3
0.07793					
## 288	0	12.890	13.12	81.89	515.9
0.06955					
## 399	0	11.060	14.83	70.31	378.2
0.07741					
## 60	0	8.618	11.79	54.34	224.5
0.09752					
## 193	0	9.720	18.22	60.73	288.1
0.06950					
## 494	0	12.460	12.83	78.83	477.3
0.07372					
## 297	0	10.910	12.35	69.14	363.7
0.08518					
## 314	0	11.540	10.72	73.73	409.1
0.08597					
## 141	0	9.738	11.97	61.24	288.5
0.09250					
## 167	0	10.800	9.71	68.77	357.6
0.09594					
## 160	0	10.900	12.96	68.69	366.8
0.07515					
## 308	0	9.000	14.40	56.36	246.3
0.07005					
##	compactness_mean	concavity_mean	concave.points_mean	symmetry_mean	
## 83	0.26650	0.3339000	0.184500	0.1829	
## 123	0.28670	0.4268000	0.201200	0.2655	

## 181	0.19140	0.2871000	0.187800	0.1800
## 203	0.20840	0.3523000	0.162000	0.2200
## 213	0.15160	0.3201000	0.159500	0.1648
## 340	0.12830	0.2308000	0.141000	0.1797
## 353	0.23630	0.3368000	0.191300	0.1956
## 462	0.19880	0.3635000	0.168900	0.2061
## 522	0.21060	0.2310000	0.147100	0.1991
## 109	0.27680	0.4264000	0.182300	0.2556
## 237	0.16820	0.1950000	0.123700	0.1909
## 565	0.11590	0.2439000	0.138900	0.1726
## 370	0.19540	0.2448000	0.150100	0.1824
## 394	0.20870	0.2810000	0.156200	0.2162
## 568	0.27700	0.3514000	0.152000	0.2397
## 504	0.12750	0.1676000	0.100300	0.1505
## 165	0.11450	0.1324000	0.097020	0.1801
## 564	0.22360	0.3174000	0.147400	0.2149
## 79	0.34540	0.3754000	0.160400	0.2906
## 266	0.11430	0.1367000	0.086460	0.1769
## 251	0.16060	0.2712000	0.131000	0.2205
## 182	0.28320	0.2487000	0.149600	0.2395
## 324	0.18750	0.2565000	0.150400	0.2569
## 273	0.19610	0.2195000	0.108800	0.1721
## 369	0.08562	0.1168000	0.084650	0.1717
## 536	0.17390	0.2085000	0.132200	0.2127
## 303	0.18380	0.2283000	0.128000	0.2249
## 84	0.17910	0.1937000	0.146900	0.1634
## 500	0.16440	0.2188000	0.112100	0.1848
## 450	0.11750	0.1572000	0.115500	0.1554
## 261	0.10880	0.1519000	0.093330	0.1814
## 24	0.10220	0.1097000	0.086320	0.1769
## 367	0.16690	0.1641000	0.126500	0.1875
## 566	0.10340	0.1440000	0.097910	0.1752
## 257	0.20630	0.1784000	0.114400	0.1893
## 3	0.15990	0.1974000	0.127900	0.2069
## 130	0.15890	0.2545000	0.114900	0.2202
## 163	0.16660	0.2508000	0.128600	0.2027
## 234	0.10740	0.1554000	0.083400	0.1448
## 433	0.14890	0.2133000	0.125900	0.1724
## 373	0.15150	0.1932000	0.125500	0.1973
## 31	0.18870	0.2319000	0.124400	0.2183
## 534	0.13130	0.1523000	0.101500	0.2166
## 19	0.10270	0.1479000	0.094980	0.1582
## 301	0.16420	0.2197000	0.106200	0.1792
## 220	0.11300	0.1145000	0.066370	0.1428
## 211	0.13480	0.1640000	0.095610	0.1765
## 344	0.13390	0.1863000	0.110300	0.2082
## 452	0.09871	0.1655000	0.090630	0.1663
## 281	0.14530	0.1921000	0.096640	0.1902
## 96	0.13130	0.1465000	0.086830	0.2095
## 390	0.13180	0.1856000	0.102100	0.1989

## 488	0.14480	0.2256000	0.119400	0.1823
## 240	0.12980	0.1417000	0.088110	0.1809
## 374	0.10760	0.1527000	0.089410	0.1571
## 5	0.13280	0.1980000	0.104300	0.1809
## 219	0.13060	0.1272000	0.086910	0.2094
## 253	0.18490	0.2417000	0.097400	0.1733
## 245	0.15580	0.2049000	0.088860	0.1978
## 33	0.14960	0.2417000	0.120300	0.2248
## 13	0.24580	0.2065000	0.111800	0.2397
## 401	0.25760	0.3189000	0.119800	0.2113
## 366	0.11310	0.0979900	0.077850	0.1618
## 518	0.13100	0.1411000	0.094310	0.1802
## 34	0.17190	0.1657000	0.075930	0.1853
## 43	0.21900	0.2107000	0.099610	0.2310
## 322	0.08564	0.1155000	0.077260	0.1928
## 169	0.16030	0.2159000	0.104300	0.1538
## 26	0.22760	0.2229000	0.140100	0.3040
## 434	0.13890	0.1594000	0.087440	0.1943
## 2	0.07864	0.0869000	0.070170	0.1812
## 255	0.11880	0.1379000	0.085910	0.1776
## 238	0.08348	0.0904200	0.060220	0.1467
## 57	0.12670	0.1323000	0.089940	0.1917
## 54	0.14850	0.1772000	0.106000	0.2092
## 447	0.13140	0.1698000	0.082930	0.1713
## 517	0.12480	0.1569000	0.094510	0.1860
## 283	0.14420	0.1626000	0.094640	0.1893
## 259	0.31140	0.3176000	0.137700	0.2495
## 46	0.16860	0.1974000	0.100900	0.1907
## 1	0.27760	0.3001000	0.147100	0.2419
## 157	0.16650	0.1855000	0.105400	0.1971
## 88	0.12060	0.1468000	0.082710	0.1953
## 71	0.10290	0.1080000	0.079510	0.1582
## 86	0.10530	0.1335000	0.087950	0.2132
## 260	0.16390	0.1751000	0.083990	0.2091
## 162	0.11850	0.1193000	0.096670	0.1741
## 25	0.14570	0.1525000	0.091700	0.1995
## 28	0.10660	0.1490000	0.077310	0.1697
## 199	0.14280	0.1114000	0.067720	0.1767
## 336	0.10560	0.1508000	0.099340	0.1727
## 563	0.20870	0.2550000	0.094290	0.2128
## 469	0.20040	0.2136000	0.100200	0.1696
## 231	0.15720	0.1910000	0.109000	0.2131
## 122	0.11000	0.1457000	0.086650	0.1966
## 73	0.18300	0.1692000	0.079440	0.1927
## 135	0.09709	0.1153000	0.068470	0.1692
## 352	0.23640	0.2914000	0.124200	0.2375
## 78	0.21460	0.1684000	0.108000	0.2152
## 499	0.13170	0.1491000	0.091830	0.1832
## 409	0.13040	0.1201000	0.088240	0.1992
## 461	0.11100	0.1007000	0.064310	0.1793

## 493	0.12890	0.1170000	0.077620	0.2116
## 128	0.08028	0.0927100	0.056270	0.1946
## 258	0.22840	0.2448000	0.124200	0.2398
## 338	0.14020	0.1060000	0.060900	0.1953
## 119	0.17520	0.2133000	0.094790	0.2096
## 214	0.11460	0.1682000	0.065970	0.1308
## 7	0.10900	0.1127000	0.074000	0.1794
## 371	0.14970	0.1811000	0.087730	0.2175
## 418	0.15710	0.1522000	0.084810	0.2085
## 318	0.11170	0.1130000	0.079500	0.1807
## 263	0.12730	0.0969700	0.075070	0.2108
## 278	0.05884	0.0802000	0.058430	0.1550
## 330	0.12830	0.1799000	0.079810	0.1869
## 18	0.20220	0.1722000	0.102800	0.2164
## 29	0.16970	0.1683000	0.087510	0.1926
## 204	0.17680	0.1558000	0.091760	0.2251
## 329	0.13190	0.1478000	0.084880	0.1948
## 265	0.08995	0.0906100	0.065270	0.1867
## 16	0.15950	0.1639000	0.073640	0.2303
## 510	0.18700	0.2030000	0.085200	0.1807
## 480	0.18930	0.2236000	0.091940	0.2151
## 254	0.10410	0.1266000	0.083530	0.1813
## 275	0.07027	0.0569900	0.047440	0.1538
## 198	0.08642	0.1103000	0.057780	0.1770
## 393	0.15620	0.1891000	0.091130	0.1929
## 442	0.11090	0.1204000	0.057360	0.1467
## 66	0.14790	0.1267000	0.090290	0.1953
## 431	0.22250	0.2733000	0.097110	0.2041
## 178	0.15560	0.1793000	0.088660	0.1794
## 567	0.10230	0.0925100	0.053020	0.1590
## 120	0.06722	0.0729300	0.055960	0.2129
## 187	0.08468	0.0816900	0.058140	0.1621
## 202	0.11980	0.1036000	0.074880	0.1506
## 132	0.12230	0.1466000	0.080870	0.1931
## 284	0.18020	0.1948000	0.090520	0.1876
## 36	0.13360	0.1348000	0.060180	0.1896
## 9	0.19320	0.1859000	0.093530	0.2350
## 354	0.09769	0.1235000	0.065530	0.1647
## 30	0.11570	0.0987500	0.079530	0.1739
## 173	0.15550	0.2032000	0.109700	0.1966
## 153	0.15990	0.4108000	0.078570	0.2548
## 63	0.20080	0.2135000	0.086530	0.1949
## 208	0.07304	0.0695000	0.053900	0.2026
## 95	0.15530	0.1700000	0.088150	0.1855
## 133	0.12840	0.1043000	0.056130	0.2160
## 4	0.28390	0.2414000	0.105200	0.2597
## 168	0.09182	0.0842200	0.065760	0.1893
## 195	0.19800	0.1697000	0.088780	0.1737
## 502	0.16810	0.1357000	0.067590	0.2275
## 58	0.13650	0.1293000	0.081230	0.2027

## 35	0.15590	0.1354000	0.077520	0.1998
## 445	0.12320	0.1090000	0.062540	0.1720
## 76	0.08424	0.0976900	0.066380	0.1798
## 118	0.16490	0.1690000	0.089230	0.2157
## 139	0.13050	0.1539000	0.086240	0.1957
## 27	0.18680	0.1425000	0.087830	0.2252
## 10	0.23960	0.2273000	0.085430	0.2030
## 92	0.10360	0.1122000	0.074830	0.1717
## 14	0.10020	0.0993800	0.053640	0.1847
## 15	0.22930	0.2128000	0.080250	0.2069
## 106	0.17650	0.2071000	0.096010	0.1925
## 224	0.12040	0.1147000	0.064620	0.1935
## 23	0.21350	0.2077000	0.097560	0.2521
## 197	0.12670	0.1385000	0.065260	0.1834
## 415	0.04605	0.0468600	0.027390	0.1852
## 262	0.06290	0.0289100	0.028370	0.1564
## 537	0.11540	0.1463000	0.061390	0.1926
## 129	0.18070	0.1138000	0.085340	0.2001
## 191	0.24130	0.1981000	0.066180	0.2384
## 65	0.12620	0.1128000	0.068730	0.1905
## 513	0.14690	0.1445000	0.081720	0.2116
## 142	0.11370	0.0944700	0.059430	0.1861
## 12	0.12920	0.0995400	0.066060	0.1842
## 215	0.13060	0.1115000	0.064620	0.2235
## 276	0.07210	0.0592900	0.074040	0.2015
## 183	0.08799	0.0659300	0.051890	0.1618
## 364	0.08468	0.0586200	0.048350	0.1495
## 230	0.17990	0.1695000	0.068610	0.2123
## 48	0.12310	0.1226000	0.073400	0.2128
## 37	0.10980	0.1319000	0.055980	0.1885
## 331	0.13710	0.1204000	0.070410	0.1782
## 8	0.16450	0.0936600	0.059850	0.2196
## 113	0.22330	0.3003000	0.077980	0.1704
## 87	0.09947	0.1204000	0.049380	0.2075
## 542	0.12300	0.1009000	0.038900	0.1872
## 386	0.06636	0.0839000	0.052710	0.1627
## 194	0.13530	0.1085000	0.045620	0.1943
## 90	0.13390	0.0996600	0.070640	0.2116
## 456	0.07426	0.0281900	0.032640	0.1375
## 561	0.11260	0.0446200	0.043040	0.1537
## 17	0.07200	0.0739500	0.052590	0.1586
## 492	0.06217	0.0444500	0.041780	0.1220
## 239	0.10390	0.1103000	0.044080	0.1342
## 200	0.12060	0.1180000	0.059800	0.1950
## 101	0.08511	0.0862500	0.044890	0.1609
## 6	0.17000	0.1578000	0.080890	0.2087
## 158	0.07223	0.0515000	0.027710	0.1844
## 39	0.05131	0.0239800	0.028990	0.1565
## 11	0.06669	0.0329900	0.033230	0.1528
## 436	0.11330	0.1126000	0.064630	0.1669

## 538	0.15520	0.0451500	0.045310	0.2131
## 509	0.06712	0.0552600	0.045630	0.1711
## 100	0.11410	0.0938800	0.058390	0.1879
## 529	0.09755	0.1010000	0.066150	0.1976
## 42	0.12180	0.1044000	0.056690	0.1895
## 376	0.14380	0.0665100	0.053970	0.1990
## 291	0.16760	0.1362000	0.066020	0.1714
## 44	0.14360	0.0984700	0.061580	0.1974
## 543	0.07214	0.0410500	0.030270	0.1840
## 496	0.08345	0.0682400	0.049510	0.1487
## 414	0.10250	0.0685900	0.038760	0.1944
## 55	0.07081	0.0525300	0.033340	0.1616
## 147	0.17000	0.1659000	0.074150	0.2678
## 151	0.07589	0.0313600	0.026450	0.2540
## 134	0.09462	0.0713500	0.059330	0.1816
## 544	0.06877	0.0298700	0.032750	0.1628
## 40	0.12550	0.1063000	0.054390	0.1720
## 77	0.10470	0.0687700	0.065560	0.2403
## 292	0.09823	0.0594000	0.048190	0.1879
## 91	0.08606	0.0310200	0.029570	0.1685
## 256	0.12790	0.0978900	0.052460	0.1908
## 397	0.11470	0.0858000	0.053810	0.1806
## 407	0.08501	0.0550000	0.045280	0.1735
## 490	0.07112	0.0364900	0.023070	0.1846
## 446	0.09218	0.0544100	0.042740	0.1820
## 45	0.10470	0.0825900	0.052520	0.1746
## 457	0.08574	0.0716000	0.020170	0.1799
## 209	0.14830	0.0870500	0.051020	0.1850
## 515	0.08597	0.0748600	0.043350	0.1561
## 454	0.09242	0.0689500	0.064950	0.1650
## 357	0.13040	0.0960300	0.056030	0.2035
## 555	0.05824	0.0619500	0.023430	0.1566
## 501	0.13640	0.0772100	0.061420	0.1668
## 185	0.10520	0.0537500	0.032630	0.1727
## 472	0.06000	0.0236700	0.023770	0.1854
## 485	0.12990	0.1191000	0.062110	0.1784
## 127	0.07862	0.0528500	0.030850	0.1761
## 559	0.13300	0.1029000	0.037360	0.1454
## 233	0.03574	0.0049670	0.006434	0.1845
## 422	0.18360	0.1450000	0.063000	0.2086
## 463	0.05223	0.0347600	0.017370	0.1707
## 289	0.11810	0.0927400	0.055880	0.2595
## 112	0.12090	0.1065000	0.060210	0.1735
## 505	0.22390	0.0973000	0.052520	0.2378
## 32	0.15160	0.1218000	0.051820	0.2301
## 69	0.14130	0.3130000	0.043750	0.2111
## 553	0.04234	0.0199700	0.014990	0.1539
## 470	0.14830	0.1020000	0.055640	0.1957
## 149	0.10210	0.0848700	0.055320	0.1724
## 82	0.15350	0.1169000	0.069870	0.1942

## 380	0.21540	0.1689000	0.063670	0.2196
## 170	0.07885	0.0260200	0.037810	0.1780
## 424	0.11470	0.0965700	0.048120	0.1848
## 226	0.07624	0.0572400	0.046030	0.2075
## 50	0.07698	0.0475100	0.033840	0.1809
## 244	0.06807	0.0469700	0.023440	0.1773
## 216	0.15170	0.0990100	0.056020	0.2106
## 264	0.05616	0.0420900	0.028470	0.1547
## 556	0.07658	0.0599900	0.027380	0.1593
## 546	0.06747	0.0297400	0.024430	0.1664
## 458	0.05205	0.0277200	0.020680	0.1619
## 341	0.11390	0.0800700	0.042230	0.1912
## 519	0.16610	0.0482500	0.053030	0.1709
## 206	0.09588	0.0755000	0.040790	0.1594
## 161	0.11410	0.0684300	0.037380	0.1993
## 377	0.16600	0.2280000	0.059410	0.2188
## 453	0.06450	0.0405500	0.019450	0.1615
## 514	0.08918	0.0822200	0.043490	0.1739
## 124	0.10990	0.0884200	0.057780	0.1856
## 560	0.10210	0.1112000	0.041050	0.1388
## 449	0.07800	0.0881700	0.029250	0.1473
## 148	0.11670	0.0905000	0.035620	0.1744
## 201	0.08087	0.0418700	0.041070	0.1979
## 497	0.13340	0.0801700	0.050740	0.1641
## 172	0.06288	0.0585800	0.034380	0.1598
## 205	0.10580	0.0800500	0.038210	0.1925
## 467	0.10890	0.1085000	0.035100	0.1562
## 459	0.05073	0.0120600	0.017620	0.1667
## 378	0.04726	0.0127100	0.011170	0.1421
## 319	0.19720	0.1975000	0.049080	0.2330
## 527	0.11380	0.0420100	0.031520	0.1723
## 20	0.08129	0.0666400	0.047810	0.1885
## 477	0.11080	0.0506300	0.030580	0.1506
## 381	0.11110	0.0790000	0.055500	0.2018
## 448	0.08890	0.0406900	0.022600	0.1893
## 166	0.05352	0.0194700	0.019390	0.1515
## 89	0.09445	0.0601500	0.037450	0.1930
## 107	0.10170	0.0707000	0.034850	0.1801
## 249	0.07234	0.0237900	0.016150	0.1897
## 507	0.11520	0.0817500	0.021660	0.2124
## 362	0.06373	0.0334400	0.024240	0.1815
## 74	0.12800	0.0778900	0.050690	0.1662
## 62	0.08963	0.0300000	0.009259	0.1828
## 94	0.08165	0.0397400	0.027800	0.1638
## 524	0.10700	0.0538500	0.037830	0.1714
## 136	0.05761	0.0471100	0.027040	0.1585
## 222	0.11920	0.0786000	0.044510	0.1962
## 280	0.07688	0.0447900	0.037110	0.2110
## 474	0.03398	0.0000000	0.000000	0.1701
## 164	0.10150	0.0537000	0.028220	0.1551

## 228	0.10960	0.0650500	0.037800	0.1881
## 486	0.15110	0.1544000	0.048460	0.2082
## 246	0.05971	0.0483100	0.030700	0.1737
## 484	0.07957	0.0454800	0.031600	0.1732
## 503	0.10850	0.0592800	0.032790	0.1943
## 520	0.11170	0.0388000	0.029950	0.2120
## 423	0.11680	0.0709700	0.044970	0.1886
## 236	0.06945	0.0146200	0.018960	0.1517
## 552	0.08194	0.0482400	0.022570	0.2030
## 487	0.06698	0.0519200	0.027910	0.1409
## 356	0.10380	0.1030000	0.043910	0.1533
## 483	0.11550	0.0578600	0.052660	0.1779
## 81	0.09362	0.0459100	0.022330	0.1842
## 435	0.07074	0.0334600	0.028770	0.1573
## 192	0.06601	0.0311200	0.028640	0.1694
## 241	0.06630	0.0470500	0.037310	0.1717
## 531	0.09713	0.0528200	0.044400	0.1598
## 416	0.08120	0.0255500	0.021790	0.2019
## 41	0.06031	0.0311000	0.020310	0.1784
## 540	0.11990	0.0925200	0.013640	0.2037
## 545	0.10180	0.0368800	0.023690	0.1620
## 395	0.09758	0.0478300	0.033260	0.1937
## 460	0.04626	0.0154100	0.010430	0.1621
## 104	0.09697	0.0615400	0.030290	0.1945
## 114	0.13030	0.0647600	0.030680	0.1922
## 229	0.07529	0.0543800	0.020360	0.1514
## 270	0.12890	0.0844800	0.028670	0.1668
## 332	0.11250	0.0710700	0.029500	0.1761
## 345	0.07281	0.0400600	0.032500	0.2009
## 508	0.10710	0.0406300	0.042680	0.1954
## 425	0.08333	0.0089340	0.019670	0.2538
## 506	0.22040	0.1188000	0.070380	0.2057
## 417	0.06159	0.0204700	0.012570	0.2025
## 116	0.07849	0.0332800	0.020080	0.1688
## 300	0.06797	0.0249500	0.018750	0.1695
## 439	0.06330	0.0134200	0.022930	0.1555
## 489	0.09263	0.0427900	0.031320	0.1853
## 140	0.11360	0.0463500	0.047960	0.1771
## 155	0.08498	0.0929300	0.034830	0.1822
## 516	0.08499	0.0430200	0.025940	0.1927
## 473	0.08549	0.0553900	0.032210	0.1687
## 466	0.12230	0.1010000	0.028330	0.1601
## 348	0.07780	0.0460800	0.035280	0.1521
## 512	0.05016	0.0341600	0.025410	0.1659
## 268	0.08259	0.0407200	0.021420	0.1635
## 532	0.09453	0.0420000	0.021570	0.1859
## 152	0.13050	0.1321000	0.021680	0.2222
## 287	0.10110	0.0657400	0.037910	0.1588
## 482	0.05326	0.0299500	0.020700	0.1579
## 97	0.07057	0.0249000	0.029410	0.1900

## 217	0.11540	0.0663600	0.031420	0.1967
## 293	0.07943	0.0615500	0.033700	0.1730
## 21	0.12700	0.0456800	0.031100	0.1967
## 396	0.05361	0.0268100	0.032510	0.1641
## 558	0.04971	0.0000000	0.000000	0.1742
## 428	0.05743	0.0361400	0.014040	0.2016
## 521	0.12250	0.0333200	0.024210	0.2197
## 80	0.09546	0.0388900	0.023150	0.1718
## 398	0.08895	0.0739000	0.040830	0.1574
## 306	0.05688	0.0197400	0.013130	0.1935
## 210	0.06230	0.0589200	0.031570	0.1359
## 70	0.05234	0.0365300	0.028640	0.1590
## 302	0.10140	0.0683000	0.030990	0.1781
## 56	0.05473	0.0303600	0.022780	0.1920
## 38	0.03766	0.0256200	0.029230	0.1467
## 438	0.05895	0.0353400	0.029440	0.1714
## 432	0.13160	0.0774100	0.027990	0.1811
## 410	0.06526	0.0321100	0.026530	0.1966
## 267	0.11470	0.0638700	0.026420	0.1922
## 437	0.07883	0.0179700	0.020900	0.1861
## 232	0.03813	0.0163300	0.003125	0.1869
## 384	0.12970	0.0589200	0.028800	0.1779
## 298	0.05914	0.0268500	0.035150	0.1619
## 243	0.13250	0.1548000	0.028540	0.2054
## 326	0.07664	0.0319300	0.021070	0.1707
## 333	0.06779	0.0050060	0.007583	0.1940
## 321	0.11110	0.0672600	0.039650	0.1743
## 372	0.06934	0.0339300	0.026570	0.1721
## 420	0.05978	0.0089550	0.010760	0.1615
## 363	0.07952	0.0268800	0.017810	0.1759
## 550	0.06602	0.0154800	0.008160	0.1976
## 406	0.07460	0.0494400	0.029320	0.1486
## 562	0.03558	0.0000000	0.000000	0.1060
## 408	0.08316	0.0612600	0.018670	0.1580
## 67	0.07773	0.0217200	0.015040	0.1717
## 105	0.08578	0.0299500	0.012010	0.2217
## 269	0.06219	0.0390000	0.016150	0.2010
## 350	0.12060	0.0117100	0.017870	0.2459
## 207	0.07232	0.0175600	0.019520	0.1934
## 188	0.06141	0.0380900	0.032390	0.1516
## 323	0.08834	0.0380000	0.034000	0.1543
## 312	0.03515	0.0144700	0.018770	0.1632
## 412	0.07804	0.0304600	0.024800	0.1714
## 491	0.05200	0.0171400	0.012610	0.1544
## 252	0.05991	0.0263800	0.020690	0.1834
## 476	0.08269	0.0583500	0.030780	0.1705
## 103	0.04038	0.0238300	0.017700	0.1739
## 177	0.12940	0.1307000	0.037160	0.1669
## 554	0.05605	0.0399600	0.012820	0.1692
## 171	0.06981	0.0398700	0.037000	0.1959

## 137	0.06095	0.0359200	0.026000	0.1339
## 290	0.05008	0.0239900	0.021730	0.2013
## 49	0.09092	0.0659200	0.027490	0.1675
## 368	0.07175	0.0439200	0.020270	0.1695
## 225	0.04994	0.0355400	0.024560	0.1496
## 465	0.05994	0.0485900	0.028700	0.1454
## 61	0.08061	0.0108400	0.012900	0.2743
## 375	0.06374	0.0255600	0.020310	0.1872
## 189	0.05562	0.0235300	0.015530	0.1718
## 156	0.06679	0.0388500	0.023310	0.1970
## 248	0.13460	0.1374000	0.039800	0.1596
## 539	0.04878	0.0000000	0.000000	0.1870
## 404	0.08836	0.0329600	0.023900	0.1735
## 250	0.07808	0.0432800	0.029290	0.1883
## 85	0.07165	0.0415100	0.018630	0.2079
## 110	0.06575	0.0513300	0.018990	0.1487
## 455	0.05430	0.0296600	0.022720	0.1799
## 440	0.05581	0.0208700	0.026520	0.1589
## 379	0.07548	0.0424900	0.024710	0.1792
## 51	0.04966	0.0165700	0.011150	0.1495
## 146	0.12960	0.0371000	0.030030	0.1995
## 533	0.07255	0.0175200	0.018800	0.1631
## 144	0.09509	0.0489400	0.030880	0.1778
## 471	0.06258	0.0294800	0.015140	0.2238
## 75	0.06829	0.0337200	0.022720	0.1720
## 339	0.07326	0.0251100	0.017750	0.1890
## 495	0.05275	0.0180000	0.012560	0.1713
## 535	0.09752	0.0526300	0.027880	0.1619
## 143	0.09486	0.0203100	0.018610	0.1645
## 441	0.11130	0.0945700	0.036130	0.1489
## 304	0.06678	0.0229700	0.017800	0.1482
## 358	0.05492	0.0150200	0.020880	0.1424
## 285	0.09580	0.1115000	0.033900	0.1432
## 327	0.05306	0.0176500	0.027330	0.1373
## 223	0.08502	0.0176800	0.019150	0.1910
## 196	0.05366	0.0387300	0.023770	0.1829
## 403	0.07899	0.0405700	0.018830	0.1874
## 138	0.06889	0.0350300	0.028750	0.1734
## 343	0.09097	0.0539700	0.033410	0.1776
## 421	0.07722	0.0548500	0.014280	0.2031
## 294	0.05642	0.0268800	0.022800	0.1875
## 383	0.10730	0.0794300	0.029780	0.1203
## 126	0.06136	0.0142000	0.011410	0.1614
## 131	0.09509	0.0285500	0.028820	0.1880
## 557	0.07504	0.0050250	0.011160	0.1791
## 279	0.04052	0.0199700	0.012380	0.1573
## 498	0.07630	0.0360900	0.023690	0.1526
## 150	0.06376	0.0288100	0.013290	0.1473
## 299	0.05220	0.0247500	0.013740	0.1635
## 59	0.03789	0.0006920	0.004167	0.1819

## 400	0.06232	0.0285300	0.016380	0.1847
## 68	0.04701	0.0370900	0.022300	0.1516
## 311	0.05253	0.0158300	0.011480	0.1936
## 530	0.09009	0.0378100	0.027980	0.1657
## 115	0.08201	0.0413200	0.019240	0.1649
## 444	0.04462	0.0199300	0.011110	0.2372
## 411	0.05313	0.0278300	0.021000	0.1601
## 315	0.05847	0.0000000	0.000000	0.2163
## 108	0.06815	0.0264300	0.019210	0.1602
## 360	0.05956	0.0271000	0.014060	0.1506
## 125	0.07325	0.0809200	0.028000	0.1422
## 479	0.08228	0.0530800	0.019690	0.1779
## 385	0.08575	0.0507700	0.028640	0.1617
## 365	0.05696	0.0218100	0.014730	0.1650
## 212	0.06900	0.0266900	0.013930	0.1533
## 478	0.05319	0.0222400	0.013390	0.1813
## 430	0.04522	0.0140200	0.018350	0.1459
## 464	0.05855	0.0336700	0.017770	0.1516
## 53	0.04751	0.0197200	0.013490	0.1868
## 481	0.07838	0.0291600	0.015270	0.1464
## 98	0.05301	0.0068290	0.007937	0.1350
## 541	0.11200	0.0673700	0.025940	0.1818
## 111	0.08404	0.0433400	0.017780	0.1584
## 296	0.06221	0.0106300	0.019170	0.1592
## 221	0.08711	0.0388800	0.025630	0.1360
## 93	0.05055	0.0326100	0.026480	0.1386
## 342	0.09228	0.0842200	0.022920	0.2036
## 451	0.10640	0.0877700	0.023860	0.1349
## 179	0.01938	0.0015950	0.001852	0.1395
## 443	0.06718	0.0105500	0.009937	0.1405
## 427	0.10130	0.0633500	0.022180	0.1925
## 335	0.04202	0.0077560	0.008535	0.1539
## 349	0.05886	0.0258700	0.023220	0.1634
## 286	0.04216	0.0018600	0.002924	0.1697
## 475	0.10690	0.0511500	0.015710	0.1861
## 247	0.04524	0.0433600	0.011050	0.1487
## 405	0.04571	0.0210900	0.020540	0.1571
## 174	0.05743	0.0236300	0.025830	0.1566
## 272	0.07608	0.0326500	0.027550	0.1769
## 305	0.07694	0.0334400	0.015020	0.1411
## 313	0.07948	0.0405200	0.025480	0.1601
## 282	0.04340	0.0224500	0.027630	0.2101
## 227	0.07722	0.0066430	0.012160	0.1788
## 548	0.08066	0.0435800	0.024380	0.1669
## 190	0.07253	0.0384400	0.016540	0.1667
## 549	0.05030	0.0233700	0.009615	0.1580
## 392	0.07428	0.0000000	0.000000	0.1985
## 274	0.04689	0.0110300	0.014070	0.2081
## 347	0.05794	0.0075100	0.008488	0.1555
## 389	0.11140	0.1007000	0.027570	0.1810

## 359	0.07698	0.0472100	0.023810	0.1930
## 295	0.08393	0.0128800	0.019240	0.1638
## 361	0.02650	0.0011940	0.005449	0.1528
## 388	0.04831	0.0204500	0.008507	0.1607
## 426	0.03912	0.0024700	0.005159	0.1630
## 218	0.05907	0.0577400	0.010710	0.1964
## 523	0.04413	0.0050670	0.005664	0.1637
## 52	0.06059	0.0185700	0.017230	0.1353
## 102	0.07568	0.0000000	0.000000	0.1930
## 271	0.02675	0.0072500	0.006250	0.1508
## 528	0.06307	0.0295800	0.026470	0.1689
## 320	0.03454	0.0134200	0.016990	0.1472
## 387	0.07823	0.0683900	0.025340	0.1646
## 117	0.12430	0.0926300	0.023080	0.1305
## 154	0.05113	0.0198200	0.017860	0.1830
## 337	0.09965	0.0373800	0.020980	0.1652
## 569	0.04362	0.0000000	0.000000	0.1587
## 547	0.04994	0.0101200	0.005495	0.1885
## 325	0.06545	0.0199400	0.016920	0.1638
## 72	0.15310	0.0860600	0.028720	0.1902
## 99	0.07525	0.0419600	0.033500	0.1620
## 551	0.04227	0.0000000	0.000000	0.1661
## 307	0.05251	0.0014610	0.003261	0.1632
## 184	0.08155	0.0618100	0.023610	0.1167
## 346	0.09159	0.0358100	0.020370	0.1633
## 413	0.06053	0.0373500	0.005128	0.1274
## 159	0.05241	0.0197200	0.019630	0.1590
## 382	0.07079	0.0354600	0.020740	0.2003
## 351	0.03630	0.0083060	0.011620	0.1671
## 525	0.08419	0.0233000	0.024160	0.1387
## 22	0.06492	0.0295600	0.020760	0.1815
## 64	0.08751	0.0598800	0.021800	0.2341
## 47	0.05943	0.0158800	0.005917	0.1769
## 277	0.03872	0.0014870	0.003333	0.1954
## 316	0.03834	0.0044730	0.006423	0.1215
## 310	0.03735	0.0045590	0.008829	0.1453
## 468	0.05428	0.0147900	0.005769	0.1680
## 419	0.05794	0.0236000	0.024020	0.1583
## 511	0.09661	0.0672600	0.026390	0.1499
## 526	0.07632	0.0256500	0.015100	0.1678
## 121	0.06685	0.0351200	0.026230	0.1667
## 391	0.07542	0.0192300	0.019680	0.1800
## 429	0.03834	0.0136900	0.013700	0.1511
## 235	0.04087	0.0165200	0.016670	0.1551
## 242	0.03393	0.0105300	0.011080	0.1546
## 186	0.04695	0.0015970	0.002404	0.1703
## 180	0.03774	0.0091930	0.013300	0.1466
## 402	0.05242	0.0260600	0.017960	0.1601
## 328	0.03892	0.0015460	0.005592	0.1382
## 175	0.04302	0.0000000	0.000000	0.1928

## 317	0.03212	0.0112300	0.005051	0.1673
## 334	0.04458	0.0009737	0.002941	0.1773
## 355	0.06064	0.0450500	0.014710	0.1690
## 309	0.03614	0.0027580	0.004419	0.1365
## 176	0.04276	0.0000000	0.000000	0.1722
## 145	0.05139	0.0225100	0.007875	0.1399
## 288	0.03729	0.0226000	0.011710	0.1337
## 399	0.04768	0.0271200	0.007246	0.1535
## 60	0.05272	0.0206100	0.007799	0.1683
## 193	0.02344	0.0000000	0.000000	0.1653
## 494	0.04043	0.0071730	0.011490	0.1613
## 297	0.04721	0.0123600	0.013690	0.1449
## 314	0.05969	0.0136700	0.008907	0.1833
## 141	0.04102	0.0000000	0.000000	0.1903
## 167	0.05736	0.0253100	0.016980	0.1381
## 160	0.03718	0.0030900	0.006588	0.1442
## 308	0.03116	0.0036810	0.003472	0.1788
##	fractal_dimension_mean CV_prediction			
## 83	0.06782	1.0000000000		
## 123	0.06877	1.0000000000		
## 181	0.05770	1.0000000000		
## 203	0.06229	1.0000000000		
## 213	0.05525	1.0000000000		
## 340	0.05506	1.0000000000		
## 353	0.06121	1.0000000000		
## 462	0.05623	1.0000000000		
## 522	0.06739	1.0000000000		
## 109	0.07039	1.0000000000		
## 237	0.06309	1.0000000000		
## 565	0.05623	0.9999999997		
## 370	0.06140	0.9999999991		
## 394	0.06606	0.9999999990		
## 568	0.07016	0.9999999988		
## 504	0.05484	0.9999999987		
## 165	0.05553	0.9999999986		
## 564	0.06879	0.9999999977		
## 79	0.08142	0.9999999976		
## 266	0.05674	0.9999999969		
## 251	0.05898	0.9999999939		
## 182	0.07398	0.9999999913		
## 324	0.06670	0.9999999852		
## 273	0.06194	0.9999999674		
## 369	0.05054	0.9999999572		
## 536	0.06251	0.9999998852		
## 303	0.07469	0.9999997887		
## 84	0.07224	0.9999997806		
## 500	0.06222	0.9999997351		
## 450	0.05661	0.9999997327		
## 261	0.05572	0.9999997041		
## 24	0.05278	0.9999996712		

## 367	0.06020	0.99999996684
## 566	0.05533	0.99999996659
## 257	0.06232	0.99999995330
## 3	0.05999	0.99999994262
## 130	0.06113	0.99999993988
## 163	0.06082	0.99999993953
## 234	0.05592	0.99999992874
## 433	0.06053	0.99999991759
## 373	0.06183	0.99999990289
## 31	0.06197	0.99999988146
## 534	0.05419	0.99999979648
## 19	0.05395	0.99999976508
## 301	0.06552	0.99999969613
## 220	0.05313	0.99999969289
## 211	0.05024	0.99999956874
## 344	0.05715	0.99999954490
## 452	0.05391	0.99999954477
## 281	0.06220	0.99999943523
## 96	0.05649	0.99999928833
## 390	0.05884	0.99999928236
## 488	0.06115	0.99999925796
## 240	0.05966	0.99999894737
## 374	0.05478	0.99999882028
## 5	0.05883	0.99999872748
## 219	0.05581	0.99999862032
## 253	0.06697	0.99999801014
## 245	0.06000	0.99999789715
## 33	0.06382	0.99999779189
## 13	0.07800	0.99999773053
## 401	0.07115	0.99999772570
## 366	0.05557	0.99999700026
## 518	0.06188	0.99999659743
## 34	0.06261	0.99999656024
## 43	0.06343	0.99999568051
## 322	0.05096	0.99999395986
## 169	0.06365	0.99999354396
## 26	0.07413	0.99999183768
## 434	0.06132	0.99999079729
## 2	0.05667	0.99998937819
## 255	0.05647	0.99998848191
## 238	0.05177	0.99998798728
## 57	0.05961	0.99998797876
## 54	0.06310	0.99998683283
## 447	0.05916	0.99998152520
## 517	0.05941	0.99998069242
## 283	0.05892	0.99998000723
## 259	0.08104	0.99997983174
## 46	0.06049	0.99997756814
## 1	0.07871	0.99996941271
## 157	0.06166	0.99996399454

## 88	0.05629	0.99995667721
## 71	0.05461	0.99995286353
## 86	0.06022	0.99994522065
## 260	0.06650	0.99994424955
## 162	0.05176	0.99993858070
## 25	0.06330	0.99993051137
## 28	0.05699	0.99992164441
## 199	0.05529	0.99991953769
## 336	0.06071	0.99990826262
## 563	0.07152	0.99988920420
## 469	0.07369	0.99988776925
## 231	0.06325	0.99988693049
## 122	0.06213	0.99988139163
## 73	0.06487	0.99987774121
## 135	0.05727	0.99986406823
## 352	0.07603	0.99986104367
## 78	0.06673	0.99983591949
## 499	0.06697	0.99982732525
## 409	0.06069	0.99979659271
## 461	0.06281	0.99976535790
## 493	0.06077	0.99972031090
## 128	0.05044	0.99961825133
## 258	0.07596	0.99961059388
## 338	0.06083	0.99960628334
## 119	0.07331	0.99959702247
## 214	0.05866	0.99958551032
## 7	0.05742	0.99956123841
## 371	0.06218	0.99952823819
## 418	0.06864	0.99948836064
## 318	0.05664	0.99948429825
## 263	0.05464	0.99943931414
## 278	0.04996	0.99939816853
## 330	0.06532	0.99917847982
## 18	0.07356	0.99915021075
## 29	0.06540	0.99887828864
## 204	0.07421	0.99866262129
## 329	0.06277	0.99857139176
## 265	0.05580	0.99853058583
## 16	0.07077	0.99839282796
## 510	0.07083	0.99783273458
## 480	0.06578	0.99768166390
## 254	0.05613	0.99736294345
## 275	0.05510	0.99721627229
## 198	0.05340	0.99720591319
## 393	0.06744	0.99683152923
## 442	0.05407	0.99666720583
## 66	0.06654	0.99656495681
## 431	0.06898	0.99557030666
## 178	0.06323	0.99504008657
## 567	0.05648	0.99447634746

## 120	0.05025 0.99434094121
## 187	0.05425 0.99424768164
## 202	0.05491 0.99343353692
## 132	0.05796 0.99312937483
## 284	0.06684 0.99263827578
## 36	0.05656 0.99258910343
## 9	0.07389 0.99234623841
## 354	0.06464 0.98916979014
## 30	0.06149 0.98702501683
## 173	0.07069 0.98684212974
## 153	0.09296 0.98601907130
## 63	0.07292 0.98598395227
## 208	0.05223 0.98556219063
## 95	0.06284 0.98326616917
## 133	0.05891 0.98323616352
## 4	0.09744 0.98171810882
## 168	0.05534 0.98153584906
## 195	0.06672 0.98142429477
## 502	0.07237 0.98009271887
## 58	0.06758 0.97789020070
## 35	0.06515 0.97751027720
## 445	0.05780 0.97726144551
## 76	0.05391 0.97678467576
## 118	0.06768 0.97333860188
## 139	0.06216 0.97150306669
## 27	0.06924 0.97108197130
## 10	0.08243 0.96354420150
## 92	0.06097 0.95893392770
## 14	0.05338 0.95862331612
## 15	0.07682 0.95593996955
## 106	0.07692 0.95516776246
## 224	0.06303 0.94737530225
## 23	0.07032 0.94283962326
## 197	0.06877 0.93811669410
## 415	0.05294 0.93692649083
## 262	0.05307 0.93615482705
## 537	0.05982 0.92728329818
## 129	0.06467 0.92417084128
## 191	0.07542 0.91901499837
## 65	0.06590 0.90682752367
## 513	0.07325 0.90024681518
## 142	0.06248 0.89742386494
## 12	0.06082 0.89581036386
## 215	0.06433 0.86565202284
## 276	0.05875 0.84286318638
## 183	0.05549 0.82857706240
## 364	0.05593 0.81495269047
## 230	0.07254 0.79875787131
## 48	0.06777 0.79374389828
## 37	0.06125 0.78740887023

## 331	0.05976	0.77991878038
## 8	0.07451	0.75392274875
## 113	0.07769	0.74728021198
## 87	0.05636	0.74681501809
## 542	0.06341	0.74464120589
## 386	0.05416	0.74087390312
## 194	0.06937	0.71021958815
## 90	0.06346	0.70989762723
## 456	0.06016	0.70770050288
## 561	0.06171	0.69852000051
## 17	0.05922	0.68834952280
## 492	0.05243	0.68335681614
## 239	0.06129	0.67206206156
## 200	0.06466	0.66995909716
## 101	0.05871	0.65657360311
## 6	0.07613	0.64519928525
## 158	0.05268	0.62772090940
## 39	0.05504	0.62274918995
## 11	0.05697	0.56879692766
## 436	0.06544	0.55804366664
## 538	0.07405	0.53539577891
## 509	0.05657	0.52813482220
## 100	0.06390	0.52799091722
## 529	0.06457	0.52131983776
## 42	0.06870	0.49725297945
## 376	0.06572	0.47801049989
## 291	0.07192	0.47232004641
## 44	0.06782	0.46885453124
## 543	0.05680	0.46784538129
## 496	0.05748	0.45823641194
## 414	0.05913	0.44920542483
## 55	0.05684	0.43312207292
## 147	0.07371	0.42661684490
## 151	0.06087	0.42327023354
## 134	0.05723	0.41061096347
## 544	0.05781	0.37349220515
## 40	0.06419	0.35480522545
## 77	0.06641	0.35442737837
## 292	0.05852	0.33338600741
## 91	0.05866	0.32831021357
## 256	0.06130	0.31891374216
## 397	0.06079	0.31605353634
## 407	0.05875	0.30300462865
## 490	0.05325	0.29796648984
## 446	0.06850	0.29048206931
## 45	0.06177	0.28072996378
## 457	0.06166	0.27849718362
## 209	0.07310	0.27422260835
## 515	0.05915	0.26139479805
## 454	0.06121	0.26137909192

## 357	0.06501	0.24986263973
## 555	0.05708	0.24557496379
## 501	0.06869	0.24513879520
## 185	0.06317	0.23028982970
## 472	0.05698	0.23007874737
## 485	0.06259	0.22864557845
## 127	0.06130	0.22634491459
## 559	0.06147	0.20515013221
## 233	0.05828	0.18251515865
## 422	0.07406	0.17965264663
## 463	0.05433	0.17950228130
## 289	0.06233	0.17202080065
## 112	0.07070	0.16899500098
## 505	0.09502	0.16361286451
## 32	0.07799	0.16126676536
## 69	0.08046	0.16026234317
## 553	0.05637	0.15807464368
## 470	0.07255	0.14591841961
## 149	0.06081	0.14269671080
## 82	0.06902	0.14147154507
## 380	0.07950	0.13426007889
## 170	0.05650	0.13292543990
## 424	0.06181	0.13008839684
## 226	0.05448	0.12653290870
## 50	0.05718	0.11734412939
## 244	0.05429	0.11638186500
## 216	0.06916	0.10961823000
## 264	0.05443	0.10927871933
## 556	0.06127	0.10707431826
## 546	0.05801	0.10501742317
## 458	0.05584	0.09785326973
## 341	0.06412	0.09743517315
## 519	0.07253	0.09512063841
## 206	0.05986	0.09473789908
## 161	0.06453	0.08957717362
## 377	0.08450	0.08837422607
## 453	0.06104	0.07634275324
## 514	0.05640	0.07480009078
## 124	0.06402	0.06960992128
## 560	0.06570	0.06958240131
## 449	0.05746	0.06370016629
## 148	0.06493	0.06223602466
## 201	0.06013	0.06134694988
## 497	0.06854	0.05852283986
## 172	0.05671	0.05472883743
## 205	0.06373	0.05220410041
## 467	0.06020	0.05203558763
## 459	0.05449	0.05063304735
## 378	0.05763	0.05037142713
## 319	0.08743	0.04693861893

## 527	0.06317 0.04593790571
## 20	0.05766 0.04557433663
## 477	0.06009 0.04463559854
## 381	0.06914 0.04405005906
## 448	0.05886 0.04399746544
## 166	0.05266 0.04323399823
## 89	0.06404 0.04312128384
## 107	0.06520 0.04222481580
## 249	0.06329 0.03912168829
## 507	0.06894 0.03837331962
## 362	0.05696 0.03769520685
## 74	0.06566 0.03614252072
## 62	0.06757 0.03539767863
## 94	0.05710 0.03493507451
## 524	0.06843 0.03374856646
## 136	0.06065 0.03323784841
## 222	0.06303 0.03225048647
## 280	0.05853 0.03162702161
## 474	0.05960 0.03110379523
## 164	0.06761 0.02983005418
## 228	0.05907 0.02910193856
## 486	0.07325 0.02877053293
## 246	0.06440 0.02836662379
## 484	0.06088 0.02812348465
## 503	0.06612 0.02680752282
## 520	0.06623 0.02633248346
## 423	0.06320 0.02595251189
## 236	0.05835 0.02578762652
## 552	0.06552 0.02545576056
## 487	0.05355 0.02521993204
## 356	0.06184 0.02517767864
## 483	0.06639 0.02502895011
## 81	0.07005 0.02493958120
## 435	0.05703 0.02478606590
## 192	0.06287 0.02350890534
## 241	0.05660 0.02337118919
## 531	0.06677 0.02277911666
## 416	0.06290 0.02270913714
## 41	0.05587 0.02266039425
## 540	0.07751 0.02251152806
## 545	0.06688 0.02144864908
## 395	0.06161 0.02052168804
## 460	0.05952 0.02022636089
## 104	0.06322 0.02014624185
## 114	0.07782 0.01924043294
## 229	0.06019 0.01882186776
## 270	0.06862 0.01878911830
## 332	0.06540 0.01850967682
## 345	0.06506 0.01755426041
## 508	0.07976 0.01746926708

## 425	0.07029	0.01742720639
## 506	0.09575	0.01722561791
## 417	0.06601	0.01716469004
## 116	0.06194	0.01715500700
## 300	0.06556	0.01653002941
## 439	0.05673	0.01613575872
## 489	0.06401	0.01555498374
## 140	0.06072	0.01547349359
## 155	0.06207	0.01525573608
## 516	0.06211	0.01500101379
## 473	0.05669	0.01494760164
## 466	0.06432	0.01426358691
## 348	0.05912	0.01406873192
## 512	0.05348	0.01347705029
## 268	0.05859	0.01320429542
## 532	0.06461	0.01276080636
## 152	0.08261	0.01247591493
## 287	0.06766	0.01247357915
## 482	0.05594	0.01241589846
## 97	0.06635	0.01162950428
## 217	0.06314	0.01150176307
## 293	0.06470	0.01099874328
## 21	0.06811	0.01098201703
## 396	0.05764	0.01063743538
## 558	0.06059	0.01055295772
## 428	0.05977	0.01053959661
## 521	0.07696	0.01041100381
## 80	0.05997	0.01039654905
## 398	0.05750	0.01023662862
## 306	0.05878	0.00993778130
## 210	0.05526	0.00975077341
## 70	0.05653	0.00960020880
## 302	0.06249	0.00952933917
## 56	0.05907	0.00946751579
## 38	0.05863	0.00929009571
## 438	0.05898	0.00926999184
## 432	0.07102	0.00911789589
## 410	0.05597	0.00901061390
## 267	0.06491	0.00886507177
## 437	0.06347	0.00855864564
## 232	0.05628	0.00842004175
## 384	0.06588	0.00837782946
## 298	0.06287	0.00812339117
## 243	0.07669	0.00780831554
## 326	0.05984	0.00764688341
## 333	0.06028	0.00744435735
## 321	0.07279	0.00696648047
## 372	0.05544	0.00670703535
## 420	0.06144	0.00662735510
## 363	0.06183	0.00659837754

## 550	0.06328	0.00658025687
## 406	0.06615	0.00611739998
## 562	0.05502	0.00610987908
## 408	0.06114	0.00610658816
## 67	0.06899	0.00600383011
## 105	0.06481	0.00588028252
## 269	0.05769	0.00576844020
## 350	0.06581	0.00573733723
## 207	0.06285	0.00566226051
## 188	0.06095	0.00557386062
## 323	0.06476	0.00548498155
## 312	0.05255	0.00541148839
## 412	0.06340	0.00529892271
## 491	0.05976	0.00521477181
## 252	0.05934	0.00515815056
## 476	0.05913	0.00507405501
## 103	0.05677	0.00502156221
## 177	0.08116	0.00495412959
## 554	0.06576	0.00488180876
## 171	0.05955	0.00486007932
## 137	0.05945	0.00485662811
## 290	0.05955	0.00476208585
## 49	0.06043	0.00457606492
## 368	0.05916	0.00444836839
## 225	0.05674	0.00439279840
## 465	0.05549	0.00424589375
## 61	0.06960	0.00423030158
## 375	0.05669	0.00421012774
## 189	0.05780	0.00413680301
## 156	0.06228	0.00413593589
## 248	0.06409	0.00409542718
## 539	0.07285	0.00403454355
## 404	0.06200	0.00400150912
## 250	0.06168	0.00394077477
## 85	0.05968	0.00383516688
## 110	0.06529	0.00383431493
## 455	0.05826	0.00378073837
## 440	0.05586	0.00365222974
## 379	0.05897	0.00360566937
## 51	0.05888	0.00354813491
## 146	0.07839	0.00353969699
## 533	0.06155	0.00352452190
## 144	0.06235	0.00343889571
## 471	0.06413	0.00341582856
## 75	0.05914	0.00340471045
## 339	0.06331	0.00331402306
## 495	0.05888	0.00329755162
## 535	0.06408	0.00325270512
## 143	0.06562	0.00322673948
## 441	0.06640	0.00298014333

## 304	0.06600	0.00295799940
## 358	0.05883	0.00294170448
## 285	0.05935	0.00282664975
## 327	0.05700	0.00281889883
## 223	0.06908	0.00281880075
## 196	0.05667	0.00281342988
## 403	0.05899	0.00280323674
## 138	0.05865	0.00280053346
## 343	0.06907	0.00279564434
## 421	0.06267	0.00277725915
## 294	0.05715	0.00275302383
## 383	0.06659	0.00271563304
## 126	0.05890	0.00271556342
## 131	0.06471	0.00265874782
## 557	0.06331	0.00264587846
## 279	0.05520	0.00263618211
## 498	0.06046	0.00260753430
## 150	0.05580	0.00259790483
## 299	0.05586	0.00248099783
## 59	0.05501	0.00245291069
## 400	0.06019	0.00244884790
## 68	0.05667	0.00221702181
## 311	0.06128	0.00221411227
## 530	0.06608	0.00218617861
## 115	0.07633	0.00207103324
## 444	0.05768	0.00203970621
## 411	0.05913	0.00191747639
## 315	0.07359	0.00189424096
## 108	0.06066	0.00186490760
## 360	0.06959	0.00174990193
## 125	0.05823	0.00172909278
## 479	0.06574	0.00161626096
## 385	0.05594	0.00158136977
## 365	0.05701	0.00152361119
## 212	0.06057	0.00148625839
## 478	0.05536	0.00140600356
## 430	0.05544	0.00140226243
## 464	0.05859	0.00139649878
## 53	0.06110	0.00131769029
## 481	0.06284	0.00131332947
## 98	0.06890	0.00130850068
## 541	0.06782	0.00127871410
## 111	0.07065	0.00126752942
## 296	0.05912	0.00120695818
## 221	0.06344	0.00117311805
## 93	0.05318	0.00115457571
## 342	0.07125	0.00114343884
## 451	0.06612	0.00113008628
## 179	0.05234	0.00110554143
## 443	0.05848	0.00106785954

## 427	0.06915	0.00106762476
## 335	0.05945	0.00094384854
## 349	0.06372	0.00094276102
## 286	0.05855	0.00092700240
## 475	0.06837	0.00089536011
## 247	0.05635	0.00087319759
## 405	0.05708	0.00086761653
## 174	0.06669	0.00084843860
## 272	0.06270	0.00083650881
## 305	0.06243	0.00081926719
## 313	0.06140	0.00080310070
## 282	0.06113	0.00079217342
## 227	0.06450	0.00078759051
## 548	0.06714	0.00078097724
## 190	0.05474	0.00077775597
## 549	0.06235	0.00075402021
## 392	0.07098	0.00075169291
## 274	0.06312	0.00073955349
## 347	0.06048	0.00073934935
## 389	0.07252	0.00072968886
## 359	0.06621	0.00072790233
## 295	0.06100	0.00072371538
## 361	0.05185	0.00071275952
## 388	0.05474	0.00071206280
## 426	0.06439	0.00070976165
## 218	0.06315	0.00070963378
## 523	0.06343	0.00069806059
## 52	0.05953	0.00069033413
## 102	0.07818	0.00068546968
## 271	0.05376	0.00066188955
## 528	0.05808	0.00064955225
## 320	0.05561	0.00064416162
## 387	0.06154	0.00062400631
## 117	0.07163	0.00056842920
## 154	0.06105	0.00055412590
## 337	0.07238	0.00055157250
## 569	0.05884	0.00054178188
## 547	0.06201	0.00053289773
## 325	0.06129	0.00052377050
## 72	0.08980	0.00051411953
## 99	0.06582	0.00050110078
## 551	0.05948	0.00047767588
## 307	0.05894	0.00046856930
## 184	0.06217	0.00045066668
## 346	0.07005	0.00044982297
## 413	0.06724	0.00044794823
## 159	0.05907	0.00043303195
## 382	0.06246	0.00041727141
## 351	0.05731	0.00039896000
## 525	0.06891	0.00038944234

```

## 22          0.06905 0.00038607838
## 64          0.06963 0.00037608776
## 47          0.06503 0.00037556188
## 277         0.05821 0.00035953486
## 316         0.05673 0.00035181279
## 310         0.05518 0.00032717011
## 468         0.06412 0.00032085158
## 419         0.06275 0.00031447074
## 511         0.06758 0.00031389579
## 526         0.07126 0.00030923502
## 121         0.06113 0.00029710032
## 391         0.06569 0.00029210342
## 429         0.06148 0.00027926014
## 235         0.06403 0.00027189802
## 242         0.05754 0.00025825699
## 186         0.06048 0.00024860541
## 180         0.06133 0.00023284878
## 402         0.05541 0.00020501623
## 328         0.06070 0.00020291124
## 175         0.05975 0.00018829244
## 317         0.05649 0.00013567190
## 334         0.06081 0.00011243840
## 355         0.06083 0.00010785200
## 309         0.05335 0.00009605277
## 176         0.06724 0.00009239687
## 145         0.05688 0.00008411288
## 288         0.05581 0.00007831944
## 399         0.06214 0.00007612266
## 60          0.07187 0.00007224763
## 193         0.06447 0.00006885554
## 494         0.06013 0.00006254942
## 297         0.06031 0.00005899537
## 314         0.06100 0.00005190078
## 141         0.06422 0.00005071479
## 167         0.06400 0.00004676403
## 160         0.05743 0.00002903715
## 308         0.06833 0.00002302831

```

#dummy variables

```

xaxis = NULL
cumMean = NULL
cumLift = NULL
meanResponse = mean(Cancer3Final$diagnosis)
meanResponse

```

```
## [1] 0.3725835
```

#initiate variables

```

xaxis[1] = 1
cumMean[1] = meanResponse

```

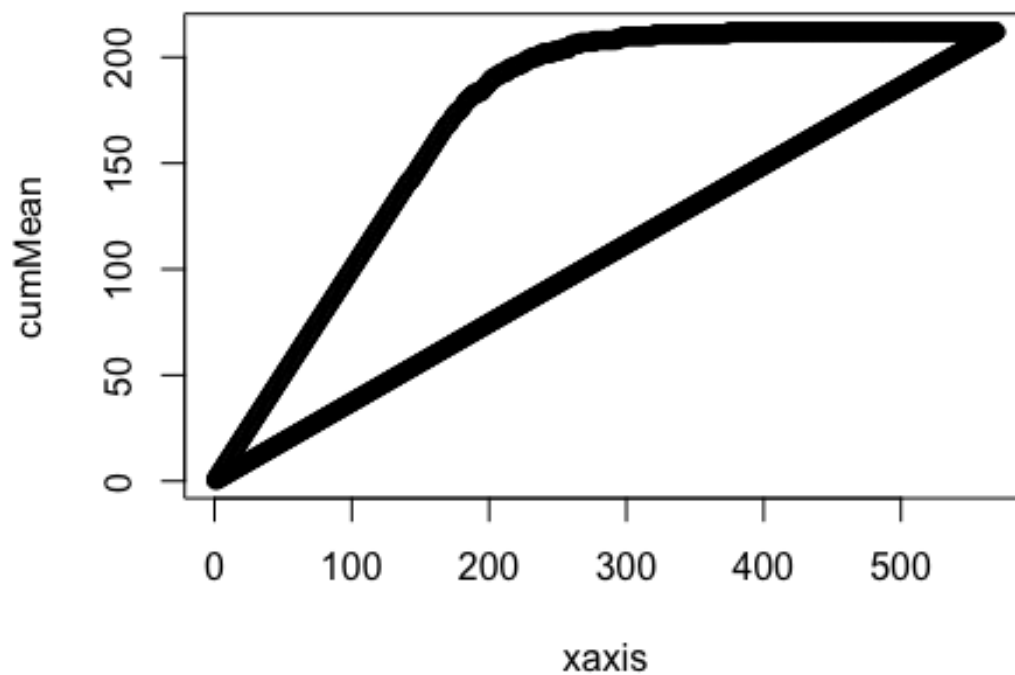
```

cumLift[1] = evalOrderedCV$diagnosis[1]

#repeating for all rows
for (i in 2 : nrow(evalOrderedCV)) {
  xaxis[i] = i
  cumMean[i] = cumMean[i-1] + meanResponse
  cumLift[i] = cumLift[i-1] + evalOrderedCV$diagnosis[i]
}

plot(cumMean ~ xaxis)
points(cumLift ~ xaxis)

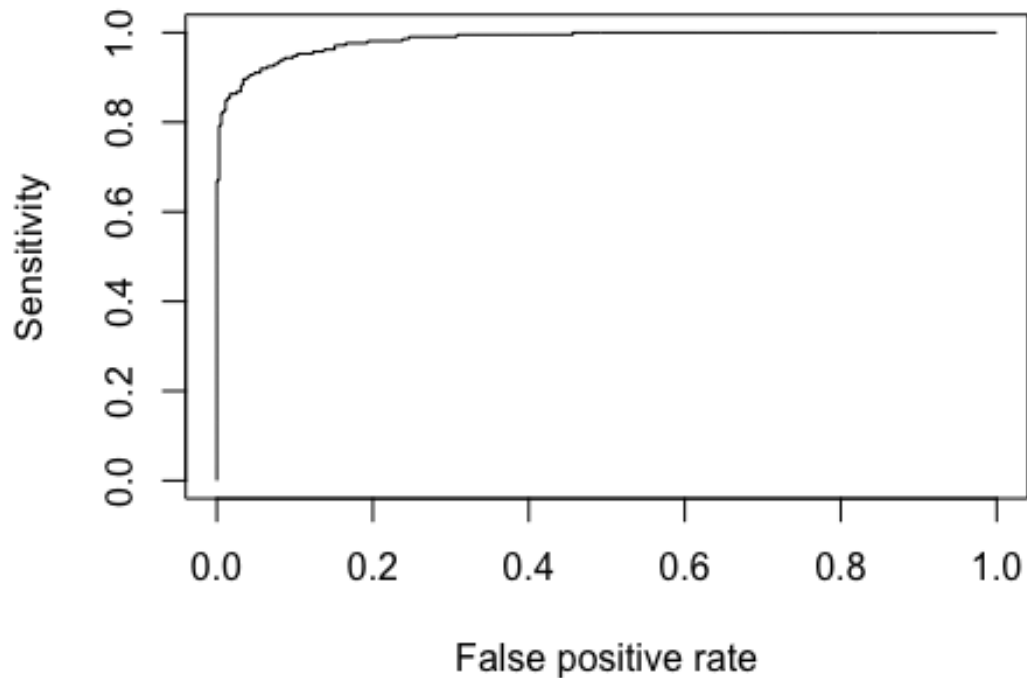
```



```

## ROC
library(ROCR)
pred_cv = prediction(Cancer3Final$CV_prediction,Cancer3Final$diagnosis)
roc_cv = performance(pred_cv,"sens","fpr")
plot(roc_cv)

```



```
falsePositive = eval[eval$falsePositive==1,]
falseNegative = eval[eval$falseNegative==1,]
falsePositive # This displays it on screen

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
smoothness_mean
## 129          0         15.10         16.39          99.58        674.5
0.1150
## 276          0         11.89         17.36          76.20        435.6
0.1225
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## 129          0.1807         0.11380         0.08534         0.2001
## 276          0.0721         0.05929         0.07404         0.2015
##      fractal_dimension_mean Prediction falsePositive falseNegative
## 129          0.06467 0.9350541          1          0
## 276          0.05875 0.7974453          1          0

falseNegative

##      diagnosis radius_mean texture_mean perimeter_mean area_mean
smoothness_mean
## 42          1         10.95         21.35          71.90        371.1
0.12270
## 40          1         13.48         20.82          88.40        559.2
```



```

0.10160
## 172      1      13.43      19.63      85.84      565.4
0.09048
## 298      1      11.76      18.14      75.00      431.1
0.09968
## 45       1      13.17      21.81      85.42      531.5
0.09714
## 206      1      15.12      16.68      98.78      716.6
0.08876
## 380      1      11.08      18.83      73.30      361.6
0.12160
## 127      1      13.61      24.69      87.76      572.6
0.09258
## 41       1      13.44      21.58      86.18      563.0
0.08162
## compactness_mean concavity_mean concave.points_mean symmetry_mean
## 42      0.12180      0.10440      0.05669      0.1895
## 40      0.12550      0.10630      0.05439      0.1720
## 172     0.06288      0.05858      0.03438      0.1598
## 298     0.05914      0.02685      0.03515      0.1619
## 45      0.10470      0.08259      0.05252      0.1746
## 206     0.09588      0.07550      0.04079      0.1594
## 380     0.21540      0.16890      0.06367      0.2196
## 127     0.07862      0.05285      0.03085      0.1761
## 41      0.06031      0.03110      0.02031      0.1784
## fractal_dimension_mean Prediction falsePositive falseNegative
## 42      0.06870 0.320989574      0      1
## 40      0.06419 0.251576794      0      1
## 172     0.05671 0.033414894      0      1
## 298     0.06287 0.006573481      0      1
## 45      0.06177 0.277845095      0      1
## 206     0.05986 0.055791303      0      1
## 380     0.07950 0.202029928      0      1
## 127     0.06130 0.160224232      0      1
## 41      0.05587 0.016167469      0      1

error = (nrow(falseNegative) + nrow(falsePositive))/nrow(eval)
error

## [1] 0.06470588

```

#Linear regression start here

```

#####
## 1. Initialize and import data
#####
set.seed(1)
options(scipen = 9999)
library(MASS)
df <- read.csv("/Users/nawwaf/Desktop/Kent/Kent Master_s/Machine
Learning/Final Project/data.csv")

```

```
Cancer3$diagnosis = as.factor(Cancer3$diagnosis) ##automatic
head(df)
```

```
##      id diagnosis radius_mean texture_mean perimeter_mean area_mean
## 1  842302      M      17.99      10.38      122.80      1001.0
## 2  842517      M      20.57      17.77      132.90      1326.0
## 3 84300903      M      19.69      21.25      130.00      1203.0
## 4 84348301      M      11.42      20.38      77.58      386.1
## 5 84358402      M      20.29      14.34      135.10      1297.0
## 6  843786      M      12.45      15.70      82.57      477.1
## smoothness_mean compactness_mean concavity_mean concave.points_mean
## 1      0.11840      0.27760      0.3001      0.14710
## 2      0.08474      0.07864      0.0869      0.07017
## 3      0.10960      0.15990      0.1974      0.12790
## 4      0.14250      0.28390      0.2414      0.10520
## 5      0.10030      0.13280      0.1980      0.10430
## 6      0.12780      0.17000      0.1578      0.08089
## symmetry_mean fractal_dimension_mean radius_se texture_se perimeter_se
## 1      0.2419      0.07871      1.0950      0.9053      8.589
## 2      0.1812      0.05667      0.5435      0.7339      3.398
## 3      0.2069      0.05999      0.7456      0.7869      4.585
## 4      0.2597      0.09744      0.4956      1.1560      3.445
## 5      0.1809      0.05883      0.7572      0.7813      5.438
## 6      0.2087      0.07613      0.3345      0.8902      2.217
## area_se smoothness_se compactness_se concavity_se concave.points_se
## 1  153.40      0.006399      0.04904      0.05373      0.01587
## 2   74.08      0.005225      0.01308      0.01860      0.01340
## 3   94.03      0.006150      0.04006      0.03832      0.02058
## 4   27.23      0.009110      0.07458      0.05661      0.01867
## 5   94.44      0.011490      0.02461      0.05688      0.01885
## 6   27.19      0.007510      0.03345      0.03672      0.01137
## symmetry_se fractal_dimension_se radius_worst texture_worst
perimeter_worst
## 1      0.03003      0.006193      25.38      17.33
184.60
## 2      0.01389      0.003532      24.99      23.41
158.80
## 3      0.02250      0.004571      23.57      25.53
152.50
## 4      0.05963      0.009208      14.91      26.50
98.87
## 5      0.01756      0.005115      22.54      16.67
152.20
## 6      0.02165      0.005082      15.47      23.75
103.40
## area_worst smoothness_worst compactness_worst concavity_worst
## 1      2019.0      0.1622      0.6656      0.7119
## 2      1956.0      0.1238      0.1866      0.2416
## 3      1709.0      0.1444      0.4245      0.4504
```

```
## 4      567.7      0.2098      0.8663      0.6869
## 5     1575.0      0.1374      0.2050      0.4000
## 6      741.6      0.1791      0.5249      0.5355
## concave.points_worst symmetry_worst fractal_dimension_worst X
## 1      0.2654      0.4601      0.11890 NA
## 2      0.1860      0.2750      0.08902 NA
## 3      0.2430      0.3613      0.08758 NA
## 4      0.2575      0.6638      0.17300 NA
## 5      0.1625      0.2364      0.07678 NA
## 6      0.1741      0.3985      0.12440 NA
```

```
df <- df[,-33]
```

```
head(df)
```

```
##      id diagnosis radius_mean texture_mean perimeter_mean area_mean
## 1  842302      M      17.99      10.38      122.80      1001.0
## 2  842517      M      20.57      17.77      132.90      1326.0
## 3 84300903      M      19.69      21.25      130.00      1203.0
## 4 84348301      M      11.42      20.38       77.58       386.1
## 5 84358402      M      20.29      14.34      135.10      1297.0
## 6  843786      M      12.45      15.70       82.57       477.1
## smoothness_mean compactness_mean concavity_mean concave.points_mean
## 1      0.11840      0.27760      0.3001      0.14710
## 2      0.08474      0.07864      0.0869      0.07017
## 3      0.10960      0.15990      0.1974      0.12790
## 4      0.14250      0.28390      0.2414      0.10520
## 5      0.10030      0.13280      0.1980      0.10430
## 6      0.12780      0.17000      0.1578      0.08089
## symmetry_mean fractal_dimension_mean radius_se texture_se perimeter_se
## 1      0.2419      0.07871      1.0950      0.9053      8.589
## 2      0.1812      0.05667      0.5435      0.7339      3.398
## 3      0.2069      0.05999      0.7456      0.7869      4.585
## 4      0.2597      0.09744      0.4956      1.1560      3.445
## 5      0.1809      0.05883      0.7572      0.7813      5.438
## 6      0.2087      0.07613      0.3345      0.8902      2.217
## area_se smoothness_se compactness_se concavity_se concave.points_se
## 1  153.40      0.006399      0.04904      0.05373      0.01587
## 2   74.08      0.005225      0.01308      0.01860      0.01340
## 3   94.03      0.006150      0.04006      0.03832      0.02058
## 4   27.23      0.009110      0.07458      0.05661      0.01867
## 5   94.44      0.011490      0.02461      0.05688      0.01885
## 6   27.19      0.007510      0.03345      0.03672      0.01137
## symmetry_se fractal_dimension_se radius_worst texture_worst
perimeter_worst
## 1      0.03003      0.006193      25.38      17.33
184.60
## 2      0.01389      0.003532      24.99      23.41
158.80
## 3      0.02250      0.004571      23.57      25.53
```

```

152.50
## 4      0.05963      0.009208      14.91      26.50
98.87
## 5      0.01756      0.005115      22.54      16.67
152.20
## 6      0.02165      0.005082      15.47      23.75
103.40
##      area_worst smoothness_worst compactness_worst concavity_worst
## 1      2019.0      0.1622      0.6656      0.7119
## 2      1956.0      0.1238      0.1866      0.2416
## 3      1709.0      0.1444      0.4245      0.4504
## 4       567.7      0.2098      0.8663      0.6869
## 5      1575.0      0.1374      0.2050      0.4000
## 6       741.6      0.1791      0.5249      0.5355
##      concave.points_worst symmetry_worst fractal_dimension_worst
## 1           0.2654           0.4601           0.11890
## 2           0.1860           0.2750           0.08902
## 3           0.2430           0.3613           0.08758
## 4           0.2575           0.6638           0.17300
## 5           0.1625           0.2364           0.07678
## 6           0.1741           0.3985           0.12440

df <- df[,-1]
#df <- na.omit(df) # Remove NA (missing) values

summary(df)

##      diagnosis      radius_mean      texture_mean      perimeter_mean
## Length:569      Min.   : 6.981      Min.   : 9.71      Min.   : 43.79
## Class :character 1st Qu.:11.700      1st Qu.:16.17      1st Qu.: 75.17
## Mode  :character Median :13.370      Median :18.84      Median : 86.24
##              Mean  :14.127      Mean  :19.29      Mean  : 91.97
##              3rd Qu.:15.780      3rd Qu.:21.80      3rd Qu.:104.10
##              Max.   :28.110      Max.   :39.28      Max.   :188.50
##      area_mean      smoothness_mean      compactness_mean      concavity_mean
## Min.   : 143.5      Min.   :0.05263      Min.   :0.01938      Min.   :0.00000
## 1st Qu.: 420.3      1st Qu.:0.08637      1st Qu.:0.06492      1st Qu.:0.02956
## Median : 551.1      Median :0.09587      Median :0.09263      Median :0.06154
## Mean   : 654.9      Mean   :0.09636      Mean   :0.10434      Mean   :0.08880
## 3rd Qu.: 782.7      3rd Qu.:0.10530      3rd Qu.:0.13040      3rd Qu.:0.13070
## Max.   :2501.0      Max.   :0.16340      Max.   :0.34540      Max.   :0.42680
##      concave.points_mean symmetry_mean      fractal_dimension_mean      radius_se
## Min.   :0.00000      Min.   :0.1060      Min.   :0.04996      Min.
## :0.1115
## 1st Qu.:0.02031      1st Qu.:0.1619      1st Qu.:0.05770      1st
## Qu.:0.2324
## Median :0.03350      Median :0.1792      Median :0.06154      Median
## :0.3242
## Mean   :0.04892      Mean   :0.1812      Mean   :0.06280      Mean
## :0.4052

```

```
## 3rd Qu.:0.07400      3rd Qu.:0.1957      3rd Qu.:0.06612      3rd
Qu.:0.4789
## Max.      :0.20120      Max.      :0.3040      Max.      :0.09744      Max.
:2.8730
## texture_se      perimeter_se      area_se      smoothness_se
## Min.      :0.3602      Min.      : 0.757      Min.      : 6.802      Min.      :0.001713
## 1st Qu.:0.8339      1st Qu.: 1.606      1st Qu.: 17.850      1st Qu.:0.005169
## Median :1.1080      Median : 2.287      Median : 24.530      Median :0.006380
## Mean      :1.2169      Mean      : 2.866      Mean      : 40.337      Mean      :0.007041
## 3rd Qu.:1.4740      3rd Qu.: 3.357      3rd Qu.: 45.190      3rd Qu.:0.008146
## Max.      :4.8850      Max.      :21.980      Max.      :542.200      Max.      :0.031130
## compactness_se      concavity_se      concave.points_se      symmetry_se
## Min.      :0.002252      Min.      :0.00000      Min.      :0.000000      Min.      :0.007882
## 1st Qu.:0.013080      1st Qu.:0.01509      1st Qu.:0.007638      1st Qu.:0.015160
## Median :0.020450      Median :0.02589      Median :0.010930      Median :0.018730
## Mean      :0.025478      Mean      :0.03189      Mean      :0.011796      Mean      :0.020542
## 3rd Qu.:0.032450      3rd Qu.:0.04205      3rd Qu.:0.014710      3rd Qu.:0.023480
## Max.      :0.135400      Max.      :0.39600      Max.      :0.052790      Max.      :0.078950
## fractal_dimension_se      radius_worst      texture_worst      perimeter_worst
## Min.      :0.0008948      Min.      : 7.93      Min.      :12.02      Min.      : 50.41
## 1st Qu.:0.0022480      1st Qu.:13.01      1st Qu.:21.08      1st Qu.: 84.11
## Median :0.0031870      Median :14.97      Median :25.41      Median : 97.66
## Mean      :0.0037949      Mean      :16.27      Mean      :25.68      Mean      :107.26
## 3rd Qu.:0.0045580      3rd Qu.:18.79      3rd Qu.:29.72      3rd Qu.:125.40
## Max.      :0.0298400      Max.      :36.04      Max.      :49.54      Max.      :251.20
## area_worst      smoothness_worst      compactness_worst      concavity_worst
## Min.      : 185.2      Min.      :0.07117      Min.      :0.02729      Min.      :0.0000
## 1st Qu.: 515.3      1st Qu.:0.11660      1st Qu.:0.14720      1st Qu.:0.1145
## Median : 686.5      Median :0.13130      Median :0.21190      Median :0.2267
## Mean      : 880.6      Mean      :0.13237      Mean      :0.25427      Mean      :0.2722
## 3rd Qu.:1084.0      3rd Qu.:0.14600      3rd Qu.:0.33910      3rd Qu.:0.3829
## Max.      :4254.0      Max.      :0.22260      Max.      :1.05800      Max.      :1.2520
## concave.points_worst      symmetry_worst      fractal_dimension_worst
## Min.      :0.00000      Min.      :0.1565      Min.      :0.05504
## 1st Qu.:0.06493      1st Qu.:0.2504      1st Qu.:0.07146
## Median :0.09993      Median :0.2822      Median :0.08004
## Mean      :0.11461      Mean      :0.2901      Mean      :0.08395
## 3rd Qu.:0.16140      3rd Qu.:0.3179      3rd Qu.:0.09208
## Max.      :0.29100      Max.      :0.6638      Max.      :0.20750
```

```
#####
## 2. Split the data in training and evaluating set (70%-30%)
#####
```

```
s = sample(length(df$diagnosis), floor(length(df$diagnosis)*0.7))
s
```

```
## [1] 129 509 471 299 270 187 307 277 494 330 37 105 485 382 326 560 422
111
## [19] 404 532 506 343 121 40 537 375 248 198 378 39 435 390 280 526 45
```

```

402
## [37] 22 193 371 499 104 555 492 465 525 176 345 110 84 29 141 252 304
287
## [55] 145 329 487 548 339 118 498 346 517 107 64 478 490 224 103 316 51
221
## [73] 290 138 569 566 443 282 143 442 285 408 170 48 204 295 24 181 476
214
## [91] 296 225 511 508 163 43 1 520 78 406 528 284 116 233 293 61 430
369
## [109] 451 86 327 522 355 563 300 49 361 500 242 440 246 305 306 247 239
219
## [127] 135 552 20 377 488 530 53 400 356 130 507 65 455 359 558 124 77
218
## [145] 98 194 19 273 418 31 419 457 174 403 237 75 16 546 265 447 92
122
## [163] 152 496 207 249 160 229 140 126 14 467 413 271 434 432 358 553 212
127
## [181] 133 41 364 373 328 36 448 297 505 495 309 391 362 344 117 379 102
50
## [199] 336 286 254 349 72 567 168 385 398 539 113 234 547 73 27 388 15
294
## [217] 62 132 35 427 429 185 459 153 420 559 255 231 28 353 148 489 538
298
## [235] 60 468 523 268 93 431 202 425 527 241 33 352 544 460 217 545 536
108
## [253] 396 565 445 209 97 502 367 437 56 342 199 256 201 414 2 363 131
466
## [271] 317 156 197 220 235 473 173 83 208 444 348 501 253 180 162 568 475
167
## [289] 454 190 161 519 446 534 338 374 25 81 222 203 529 3 179 147 279
337
## [307] 260 205 335 482 389 106 351 463 23 157 323 257 164 383 119 99 283
477
## [325] 150 410 169 322 211 318 18 44 347 472 91 251 32 223 438 551 314
308
## [343] 125 370 89 250 149 21 79 115 101 192 216 366 71 128 59 90 158
58
## [361] 421 114 381 134 34 561 354 17 340 516 535 365 66 450 244 136 392
449
## [379] 439 5 74 441 183 274 334 462 303 380 261 69 210 412 278 550 324
42
## [397] 123 259

```

```

train = df[s,]
eval = df[-s,]

```

```

#####
## 3. Run a standard linear regression model on the radius_mean of tumor
(radius_mean)
#####

```

```

reg = lm(train$radius_mean~.,data=train)
reg

##
## Call:
## lm(formula = train$radius_mean ~ ., data = train)
##
## Coefficients:
##          (Intercept)          diagnosisM          texture_mean
##          0.4606835          -0.0137459          0.0006519
##          perimeter_mean          area_mean          smoothness_mean
##          0.1365224          0.0007367          1.3821097
##          compactness_mean          concavity_mean          concave.points_mean
##          -3.8591198          -1.7216037          0.5914781
##          symmetry_mean          fractal_dimension_mean          radius_se
##          0.2746640          2.0976799          0.1718394
##          texture_se          perimeter_se          area_se
##          0.0217702          -0.0474092          0.0001944
##          smoothness_se          compactness_se          concavity_se
##          0.9108659          -0.9559376          0.9561953
##          concave.points_se          symmetry_se          fractal_dimension_se
##          3.7520342          -0.9726007          4.1570796
##          radius_worst          texture_worst          perimeter_worst
##          0.1403176          -0.0027180          -0.0079314
##          area_worst          smoothness_worst          compactness_worst
##          -0.0005058          -0.7453486          0.3788198
##          concavity_worst          concave.points_worst          symmetry_worst
##          0.0697517          -0.0377612          -0.0818514
##          fractal_dimension_worst
##          -1.1535704

summary(reg)

##
## Call:
## lm(formula = train$radius_mean ~ ., data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.287012 -0.029219  0.000602  0.025323  0.225395
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)   0.4606835   0.1228599   3.750    0.000206
## ***
## diagnosisM    -0.0137459   0.0128775  -1.067    0.286477
## texture_mean   0.0006519   0.0024728   0.264    0.792207
## perimeter_mean 0.1365224   0.0030434  44.858 < 0.00000000000000002
## ***
## area_mean     0.0007367   0.0001655   4.452    0.0000112872471560

```

```

***
## smoothness_mean      1.3821097  0.5738891  2.408      0.016518
*
## compactness_mean     -3.8591198  0.3380225 -11.417 < 0.0000000000000002
***
## concavity_mean       -1.7216037  0.3062194  -5.622    0.0000000374043744
***
## concave.points_mean   0.5914781  0.5940163  0.996      0.320039
## symmetry_mean         0.2746640  0.2197782  1.250      0.212194
## fractal_dimension_mean 2.0976799  1.6023851  1.309      0.191320
## radius_se             0.1718394  0.1094082  1.571      0.117131
## texture_se            0.0217702  0.0110654  1.967      0.049889
*
## perimeter_se         -0.0474092  0.0145597  -3.256      0.001234
**
## area_se              0.0001944  0.0004163  0.467      0.640694
## smoothness_se        0.9108659  1.9749130  0.461      0.644915
## compactness_se       -0.9559376  0.7194185 -1.329      0.184751
## concavity_se         0.9561953  0.3722864  2.568      0.010611
*
## concave.points_se    3.7520342  1.6079300  2.333      0.020164
*
## symmetry_se          -0.9726007  0.7907682 -1.230      0.219506
## fractal_dimension_se  4.1570796  3.6439049  1.141      0.254684
## radius_worst         0.1403176  0.0180395  7.778    0.00000000000000752
***
## texture_worst        -0.0027180  0.0021562 -1.261      0.208269
## perimeter_worst      -0.0079314  0.0018784 -4.223    0.0000305009984407
***
## area_worst           -0.0005058  0.0001059 -4.777    0.0000025810899927
***
## smoothness_worst     -0.7453486  0.4166172 -1.789      0.074431
.
## compactness_worst     0.3788198  0.1131446  3.348      0.000898
***
## concavity_worst       0.0697517  0.0778701  0.896      0.370976
## concave.points_worst -0.0377612  0.2693871 -0.140      0.888599
## symmetry_worst        -0.0818514  0.1410668 -0.580      0.562115
## fractal_dimension_worst -1.1535704  0.7114395 -1.621      0.105778
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05714 on 367 degrees of freedom
## Multiple R-squared:  0.9998, Adjusted R-squared:  0.9997
## F-statistic: 5.264e+04 on 30 and 367 DF,  p-value: < 0.0000000000000022

```

Task 5. predictions and test the quality of the model

```

predict = predict(reg, eval)
predict

```


##	4	6	7	8	9	10	11
12							
##	11.258443	12.455825	18.279689	13.699955	13.008323	12.416709	15.957941
15.876853							
##	13	26	30	38	46	47	52
54							
##	19.331615	17.185624	17.547557	13.036548	18.677496	8.212742	13.600788
18.148217							
##	55	57	63	67	68	70	76
80							
##	15.101735	19.169047	14.323034	9.432796	11.291802	12.746291	16.071511
12.882061							
##	82	85	87	88	94	95	96
100							
##	13.304931	12.043227	14.427302	18.826697	13.458326	15.132697	20.210401
14.451792							
##	109	112	120	137	139	142	144
146							
##	22.167256	12.560633	17.704484	11.754559	14.762580	16.127280	12.907333
11.980102							
##	151	154	155	159	165	166	171
172							
##	13.024770	11.180371	13.141122	12.079941	23.241561	14.908716	12.341570
13.443493							
##	175	177	178	182	184	186	188
189							
##	10.679539	9.930216	16.427082	21.075127	11.481954	10.110439	11.706105
11.821634							
##	191	195	196	200	206	213	215
226							
##	13.946580	14.792518	12.901295	14.476691	15.185573	28.015965	14.106910
14.363420							
##	227	228	230	232	236	238	240
243							
##	10.457696	14.941748	12.732410	11.286864	14.009496	20.403433	17.419691
11.222725							
##	245	258	262	263	264	266	267
269							
##	19.377072	15.222748	17.254746	17.398894	15.584945	20.554350	10.626638
12.869312							
##	272	275	276	281	288	289	291
292							
##	11.305167	17.855913	11.918578	19.029649	12.874824	11.348666	14.567051
14.920800							
##	301	302	310	311	312	313	315
319							
##	19.538953	12.390683	13.050941	11.693095	14.597486	12.814408	8.581425
8.837469							
##	320	321	325	331	332	333	341
350							

```

## 12.414510 10.334880 12.189118 16.120923 12.928873 11.243642 14.464112
11.909830
##      357      360      368      372      376      384      386
387
## 12.995098  9.533507 12.221893 15.176114 16.164202 12.366149 14.554218
12.210152
##      393      394      395      397      399      401      405
407
## 15.478021 21.562568 12.082324 13.529300 11.122796 18.074984 12.310115
16.133232
##      409      411      415      416      417      423      424
426
## 17.996711 11.383372 15.081545 11.893859  9.453092 11.604913 13.695094
10.051740
##      428      433      436      452      453      456      458
461
## 10.847864 20.189704 13.961869 19.519699 11.983722 13.421428 13.168375
17.016845
##      464      469      470      474      479      480      481
483
## 11.570747 17.776915 11.632106 12.225413 11.497922 16.282985 12.163525
13.445316
##      484      486      491      493      497      503      504
510
## 13.650102 12.426724 12.267783 18.006102 12.640492 12.533122 23.056426
15.433387
##      512      513      514      515      518      521      524
531
## 14.768120 13.378432 14.585072 15.040180 19.877391  9.307927 13.662207
11.759515
##      533      540      541      542      543      549      554
556
## 13.712274  7.492166 11.542220 14.536397 14.724631  9.689779  9.331060
10.266371
##      557      562      564
## 10.160189 11.175933 20.891053

me = mean(eval$radius_mean-predict)
me

## [1] 0.01130962

rmse = sqrt(mean((eval$radius_mean-predict)**2))
rmse

## [1] 0.06690827

mape = mean(abs(eval$radius_mean-predict)/eval$radius_mean)*100
mape

## [1] 0.3261073

```

```
#####
## 4. Investigate reducing complexity by dropping variables
#####
```

```
drop1(reg)
```

```
## Single term deletions
```

```
##
```

```
## Model:
```

```
## train$radius_mean ~ diagnosis + texture_mean + perimeter_mean +
##   area_mean + smoothness_mean + compactness_mean + concavity_mean +
##   concave.points_mean + symmetry_mean + fractal_dimension_mean +
##   radius_se + texture_se + perimeter_se + area_se + smoothness_se +
##   compactness_se + concavity_se + concave.points_se + symmetry_se +
##   fractal_dimension_se + radius_worst + texture_worst + perimeter_worst
+
##   area_worst + smoothness_worst + compactness_worst + concavity_worst +
##   concave.points_worst + symmetry_worst + fractal_dimension_worst
```

	Df	Sum of Sq	RSS	AIC
## <none>			1.1983	-2248.6
## diagnosis	1	0.0037	1.2020	-2249.4
## texture_mean	1	0.0002	1.1985	-2250.6
## perimeter_mean	1	6.5699	7.7682	-1506.7
## area_mean	1	0.0647	1.2630	-2229.7
## smoothness_mean	1	0.0189	1.2172	-2244.4
## compactness_mean	1	0.4256	1.6238	-2129.7
## concavity_mean	1	0.1032	1.3015	-2217.7
## concave.points_mean	1	0.0032	1.2015	-2249.6
## symmetry_mean	1	0.0051	1.2033	-2248.9
## fractal_dimension_mean	1	0.0056	1.2038	-2248.8
## radius_se	1	0.0081	1.2063	-2248.0
## texture_se	1	0.0126	1.2109	-2246.4
## perimeter_se	1	0.0346	1.2329	-2239.3
## area_se	1	0.0007	1.1990	-2250.4
## smoothness_se	1	0.0007	1.1989	-2250.4
## compactness_se	1	0.0058	1.2040	-2248.7
## concavity_se	1	0.0215	1.2198	-2243.5
## concave.points_se	1	0.0178	1.2160	-2244.8
## symmetry_se	1	0.0049	1.2032	-2249.0
## fractal_dimension_se	1	0.0042	1.2025	-2249.2
## radius_worst	1	0.1975	1.3958	-2189.9
## texture_worst	1	0.0052	1.2034	-2248.9
## perimeter_worst	1	0.0582	1.2565	-2231.7
## area_worst	1	0.0745	1.2727	-2226.6
## smoothness_worst	1	0.0105	1.2087	-2247.2
## compactness_worst	1	0.0366	1.2349	-2238.7
## concavity_worst	1	0.0026	1.2009	-2249.8
## concave.points_worst	1	0.0001	1.1983	-2250.6
## symmetry_worst	1	0.0011	1.1993	-2250.3
## fractal_dimension_worst	1	0.0086	1.2068	-2247.8

```

reg1 = lm(radius_mean~.-texture_mean, data=train)
drop1(reg1)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean
##
##              Df Sum of Sq    RSS    AIC
## <none>                1.1985 -2250.6
## diagnosis             1    0.0036 1.2021 -2251.3
## perimeter_mean        1    6.8213 8.0198 -1496.0
## area_mean             1    0.0647 1.2631 -2231.6
## smoothness_mean       1    0.0193 1.2178 -2246.2
## compactness_mean      1    0.4272 1.6256 -2131.2
## concavity_mean        1    0.1042 1.3027 -2219.4
## concave.points_mean   1    0.0033 1.2018 -2251.4
## symmetry_mean         1    0.0054 1.2039 -2250.8
## fractal_dimension_mean 1    0.0056 1.2041 -2250.7
## radius_se             1    0.0089 1.2074 -2249.6
## texture_se            1    0.0155 1.2140 -2247.4
## perimeter_se          1    0.0349 1.2333 -2241.1
## area_se               1    0.0006 1.1991 -2252.3
## smoothness_se         1    0.0008 1.1993 -2252.3
## compactness_se        1    0.0058 1.2043 -2250.6
## concavity_se          1    0.0216 1.2201 -2245.4
## concave.points_se     1    0.0179 1.2164 -2246.7
## symmetry_se           1    0.0047 1.2032 -2251.0
## fractal_dimension_se  1    0.0042 1.2027 -2251.2
## radius_worst          1    0.2016 1.4001 -2190.7
## texture_worst         1    0.0261 1.2246 -2244.0
## perimeter_worst       1    0.0583 1.2568 -2233.7
## area_worst            1    0.0749 1.2734 -2228.4
## smoothness_worst      1    0.0114 1.2099 -2248.8
## compactness_worst     1    0.0368 1.2353 -2240.5
## concavity_worst       1    0.0027 1.2012 -2251.7
## concave.points_worst  1    0.0001 1.1986 -2252.5
## symmetry_worst        1    0.0015 1.2000 -2252.1
## fractal_dimension_worst 1    0.0085 1.2069 -2249.8

```

```

reg2 = lm(radius_mean~.-texture_mean-concave.points_worst, data=train)
drop1(reg2)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst
##
##              Df Sum of Sq    RSS    AIC
## <none>                1.1986 -2252.5
## diagnosis             1    0.0037 1.2022 -2253.3
## perimeter_mean        1    6.9624 8.1610 -1491.1
## area_mean             1    0.0647 1.2633 -2233.6
## smoothness_mean       1    0.0221 1.2207 -2247.2
## compactness_mean      1    0.4296 1.6281 -2132.6
## concavity_mean        1    0.1091 1.3076 -2219.8
## concave.points_mean   1    0.0039 1.2024 -2253.2
## symmetry_mean         1    0.0055 1.2040 -2252.7
## fractal_dimension_mean 1    0.0056 1.2042 -2252.7
## radius_se             1    0.0091 1.2077 -2251.5
## texture_se            1    0.0155 1.2140 -2249.4
## perimeter_se          1    0.0348 1.2334 -2243.1
## area_se               1    0.0005 1.1991 -2254.3
## smoothness_se         1    0.0010 1.1996 -2254.2
## compactness_se        1    0.0057 1.2043 -2252.6
## concavity_se          1    0.0224 1.2209 -2247.2
## concave.points_se     1    0.0266 1.2252 -2245.8
## symmetry_se           1    0.0047 1.2032 -2253.0
## fractal_dimension_se  1    0.0042 1.2028 -2253.1
## radius_worst          1    0.2022 1.4007 -2192.5
## texture_worst         1    0.0260 1.2246 -2246.0
## perimeter_worst       1    0.0594 1.2580 -2235.3
## area_worst            1    0.0753 1.2738 -2230.3
## smoothness_worst      1    0.0137 1.2123 -2250.0
## compactness_worst     1    0.0367 1.2353 -2242.5
## concavity_worst       1    0.0028 1.2013 -2253.6
## symmetry_worst        1    0.0016 1.2002 -2254.0
## fractal_dimension_worst 1    0.0084 1.2070 -2251.7

```

```

reg3 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se,
data=train)
drop1(reg3)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
## smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
## symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
## perimeter_se + area_se + smoothness_se + compactness_se +
## concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
## radius_worst + texture_worst + perimeter_worst + area_worst +
## smoothness_worst + compactness_worst + concavity_worst +
## concave.points_worst + symmetry_worst + fractal_dimension_worst) -
## texture_mean - concave.points_worst - area_se
##
##           Df Sum of Sq    RSS    AIC
## <none>                        1.1991 -2254.3
## diagnosis                1    0.0038  1.2029 -2255.1
## perimeter_mean           1    9.0541 10.2532 -1402.2
## area_mean                1    0.0702  1.2693 -2233.7
## smoothness_mean         1    0.0224  1.2215 -2249.0
## compactness_mean        1    0.4411  1.6403 -2131.7
## concavity_mean          1    0.1086  1.3077 -2221.8
## concave.points_mean     1    0.0036  1.2027 -2255.1
## symmetry_mean           1    0.0051  1.2042 -2254.7
## fractal_dimension_mean  1    0.0053  1.2044 -2254.6
## radius_se               1    0.0140  1.2131 -2251.7
## texture_se              1    0.0150  1.2141 -2251.4
## perimeter_se            1    0.0343  1.2334 -2245.1
## smoothness_se           1    0.0009  1.2000 -2256.1
## compactness_se          1    0.0054  1.2045 -2254.6
## concavity_se            1    0.0224  1.2215 -2249.0
## concave.points_se       1    0.0263  1.2254 -2247.7
## symmetry_se             1    0.0057  1.2048 -2254.5
## fractal_dimension_se    1    0.0039  1.2030 -2255.1
## radius_worst            1    0.2536  1.4527 -2180.0
## texture_worst           1    0.0258  1.2249 -2247.9
## perimeter_worst         1    0.0607  1.2598 -2236.7
## area_worst              1    0.1164  1.3155 -2219.5
## smoothness_worst        1    0.0133  1.2124 -2252.0
## compactness_worst       1    0.0364  1.2355 -2244.4
## concavity_worst         1    0.0026  1.2017 -2255.5
## symmetry_worst          1    0.0013  1.2004 -2255.9
## fractal_dimension_worst 1    0.0080  1.2071 -2253.7

```

```

reg4 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se, data=train)
drop1(reg4)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
## smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
## symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
## perimeter_se + area_se + smoothness_se + compactness_se +
## concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
## radius_worst + texture_worst + perimeter_worst + area_worst +
## smoothness_worst + compactness_worst + concavity_worst +
## concave.points_worst + symmetry_worst + fractal_dimension_worst) -
## texture_mean - concave.points_worst - area_se - smoothness_se
##
##          Df Sum of Sq    RSS    AIC
## <none>                  1.2000 -2256.1
## diagnosis             1    0.0034  1.2034 -2256.9
## perimeter_mean        1    9.0623 10.2623 -1403.9
## area_mean             1    0.0718  1.2718 -2234.9
## smoothness_mean       1    0.0216  1.2216 -2251.0
## compactness_mean      1    0.4412  1.6412 -2133.4
## concavity_mean        1    0.1078  1.3078 -2223.8
## concave.points_mean   1    0.0032  1.2032 -2257.0
## symmetry_mean         1    0.0055  1.2055 -2256.2
## fractal_dimension_mean 1    0.0054  1.2054 -2256.3
## radius_se             1    0.0156  1.2156 -2252.9
## texture_se            1    0.0153  1.2153 -2253.0
## perimeter_se          1    0.0370  1.2370 -2245.9
## compactness_se        1    0.0045  1.2045 -2256.6
## concavity_se          1    0.0215  1.2215 -2251.0
## concave.points_se     1    0.0287  1.2287 -2248.6
## symmetry_se           1    0.0049  1.2049 -2256.4
## fractal_dimension_se  1    0.0040  1.2040 -2256.7
## radius_worst          1    0.2539  1.4539 -2181.7
## texture_worst         1    0.0259  1.2259 -2249.5
## perimeter_worst       1    0.0598  1.2598 -2238.7
## area_worst            1    0.1170  1.3170 -2221.0
## smoothness_worst      1    0.0152  1.2151 -2253.1
## compactness_worst     1    0.0357  1.2357 -2246.4
## concavity_worst       1    0.0024  1.2024 -2257.3
## symmetry_worst        1    0.0019  1.2018 -2257.4
## fractal_dimension_worst 1    0.0082  1.2081 -2255.3

reg5 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst, data=train)
drop1(reg5)

```

```

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
## smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
## symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
## perimeter_se + area_se + smoothness_se + compactness_se +
## concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
## radius_worst + texture_worst + perimeter_worst + area_worst +
## smoothness_worst + compactness_worst + concavity_worst +
## concave.points_worst + symmetry_worst + fractal_dimension_worst) -
## texture_mean - concave.points_worst - area_se - smoothness_se -
## symmetry_worst
##
##              Df Sum of Sq      RSS      AIC
## <none>              1.2018 -2257.4
## diagnosis           1    0.0036  1.2054 -2258.2
## perimeter_mean      1    9.2161 10.4180 -1399.9
## area_mean           1    0.0708  1.2726 -2236.7
## smoothness_mean     1    0.0226  1.2244 -2252.0
## compactness_mean    1    0.4397  1.6415 -2135.3
## concavity_mean      1    0.1064  1.3082 -2225.7
## concave.points_mean 1    0.0029  1.2048 -2258.5
## symmetry_mean       1    0.0037  1.2056 -2258.2
## fractal_dimension_mean 1    0.0054  1.2073 -2257.6
## radius_se           1    0.0165  1.2183 -2254.0
## texture_se          1    0.0203  1.2222 -2252.8
## perimeter_se        1    0.0370  1.2388 -2247.4
## compactness_se      1    0.0037  1.2055 -2258.2
## concavity_se        1    0.0199  1.2217 -2252.9
## concave.points_se   1    0.0312  1.2330 -2249.2
## symmetry_se         1    0.0183  1.2202 -2253.4
## fractal_dimension_se 1    0.0045  1.2064 -2257.9
## radius_worst        1    0.2521  1.4539 -2183.7
## texture_worst       1    0.0291  1.2310 -2249.9
## perimeter_worst     1    0.0612  1.2630 -2239.7
## area_worst          1    0.1152  1.3170 -2223.0
## smoothness_worst    1    0.0168  1.2187 -2253.9
## compactness_worst   1    0.0338  1.2357 -2248.4
## concavity_worst     1    0.0025  1.2043 -2258.6
## fractal_dimension_worst 1    0.0088  1.2107 -2256.5

reg6 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst, data=train)
drop1(reg6)

## Single term deletions
##
## Model:

```



```
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst
```

	Df	Sum of Sq	RSS	AIC
## <none>			1.2043	-2258.6
## diagnosis	1	0.0032	1.2075	-2259.6
## perimeter_mean	1	9.2217	10.4260	-1401.6
## area_mean	1	0.0696	1.2740	-2238.2
## smoothness_mean	1	0.0237	1.2280	-2252.8
## compactness_mean	1	0.4874	1.6917	-2125.4
## concavity_mean	1	0.1250	1.3294	-2221.3
## concave.points_mean	1	0.0020	1.2064	-2259.9
## symmetry_mean	1	0.0037	1.2080	-2259.4
## fractal_dimension_mean	1	0.0051	1.2095	-2258.9
## radius_se	1	0.0155	1.2198	-2255.5
## texture_se	1	0.0195	1.2239	-2254.2
## perimeter_se	1	0.0362	1.2405	-2248.8
## compactness_se	1	0.0040	1.2084	-2259.3
## concavity_se	1	0.0260	1.2304	-2252.1
## concave.points_se	1	0.0328	1.2372	-2249.9
## symmetry_se	1	0.0186	1.2230	-2254.5
## fractal_dimension_se	1	0.0030	1.2074	-2259.6
## radius_worst	1	0.2549	1.4592	-2184.2
## texture_worst	1	0.0286	1.2330	-2251.2
## perimeter_worst	1	0.0614	1.2657	-2240.8
## area_worst	1	0.1151	1.3194	-2224.3
## smoothness_worst	1	0.0167	1.2211	-2255.1
## compactness_worst	1	0.0494	1.2538	-2244.6
## fractal_dimension_worst	1	0.0071	1.2114	-2258.3

```
reg7 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean, data=train)
drop1(reg7)
```

```
## Single term deletions
```

```
##
```

```
## Model:
```

```
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
```

```
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean
##              Df Sum of Sq      RSS      AIC
## <none>                                1.2064 -2259.9
## diagnosis                1      0.0030  1.2094 -2260.9
## perimeter_mean           1     9.5720 10.7783 -1390.3
## area_mean                1      0.0805  1.2869 -2236.2
## smoothness_mean          1      0.0286  1.2350 -2252.6
## compactness_mean         1      0.4872  1.6936 -2126.9
## concavity_mean           1      0.2370  1.4434 -2190.5
## symmetry_mean            1      0.0040  1.2104 -2260.6
## fractal_dimension_mean   1      0.0053  1.2116 -2260.2
## radius_se                1      0.0153  1.2217 -2256.9
## texture_se               1      0.0201  1.2264 -2255.4
## perimeter_se             1      0.0371  1.2434 -2249.9
## compactness_se           1      0.0037  1.2101 -2260.7
## concavity_se             1      0.0281  1.2345 -2252.8
## concave.points_se        1      0.0519  1.2582 -2245.2
## symmetry_se              1      0.0184  1.2248 -2255.9
## fractal_dimension_se     1      0.0030  1.2094 -2260.9
## radius_worst             1      0.2697  1.4761 -2181.6
## texture_worst            1      0.0298  1.2362 -2252.2
## perimeter_worst          1      0.0600  1.2664 -2242.6
## area_worst               1      0.1301  1.3365 -2221.2
## smoothness_worst         1      0.0170  1.2233 -2256.4
## compactness_worst        1      0.0474  1.2538 -2246.6
## fractal_dimension_worst  1      0.0071  1.2135 -2259.6
```

```
reg8 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis,
data=train)
drop1(reg8)
```

```
## Single term deletions
```

```
##
```

```
## Model:
```

```
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
```

```

##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean -
##      diagnosis
##
##              Df Sum of Sq      RSS      AIC
## <none>              1.2094 -2260.9
## perimeter_mean      1    9.7089 10.9182 -1387.2
## area_mean           1    0.0785  1.2878 -2237.9
## smoothness_mean     1    0.0282  1.2376 -2253.8
## compactness_mean    1    0.4877  1.6970 -2128.1
## concavity_mean      1    0.2773  1.4867 -2180.8
## symmetry_mean       1    0.0037  1.2131 -2261.7
## fractal_dimension_mean 1    0.0052  1.2146 -2261.2
## radius_se           1    0.0143  1.2236 -2258.3
## texture_se          1    0.0215  1.2308 -2255.9
## perimeter_se        1    0.0359  1.2452 -2251.3
## compactness_se      1    0.0039  1.2132 -2261.7
## concavity_se        1    0.0352  1.2445 -2251.5
## concave.points_se   1    0.0489  1.2583 -2247.2
## symmetry_se         1    0.0206  1.2299 -2256.2
## fractal_dimension_se 1    0.0032  1.2126 -2261.9
## radius_worst        1    0.2678  1.4771 -2183.3
## texture_worst       1    0.0351  1.2444 -2251.6
## perimeter_worst     1    0.0607  1.2700 -2243.5
## area_worst          1    0.1277  1.3371 -2223.0
## smoothness_worst    1    0.0178  1.2272 -2257.1
## compactness_worst   1    0.0466  1.2559 -2247.9
## fractal_dimension_worst 1    0.0079  1.2173 -2260.4

reg9 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis-
fractal_dimension_se, data=train)
drop1(reg9)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
##      concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
##      +
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean -

```

```

##      diagnosis - fractal_dimension_se
##              Df Sum of Sq      RSS      AIC
## <none>                                1.2126 -2261.9
## perimeter_mean      1      9.7123 10.9249 -1389.0
## area_mean           1      0.0783  1.2909 -2239.0
## smoothness_mean     1      0.0274  1.2400 -2255.0
## compactness_mean    1      0.4930  1.7056 -2128.1
## concavity_mean      1      0.2809  1.4935 -2181.0
## symmetry_mean       1      0.0042  1.2168 -2262.5
## fractal_dimension_mean 1      0.0070  1.2196 -2261.6
## radius_se           1      0.0209  1.2335 -2257.1
## texture_se          1      0.0231  1.2357 -2256.4
## perimeter_se        1      0.0480  1.2606 -2248.4
## compactness_se      1      0.0012  1.2138 -2263.5
## concavity_se        1      0.0375  1.2501 -2251.8
## concave.points_se   1      0.0541  1.2667 -2246.5
## symmetry_se         1      0.0226  1.2352 -2256.6
## radius_worst        1      0.2653  1.4779 -2185.2
## texture_worst       1      0.0357  1.2483 -2252.3
## perimeter_worst     1      0.0574  1.2700 -2245.5
## area_worst          1      0.1283  1.3409 -2223.8
## smoothness_worst    1      0.0197  1.2323 -2257.5
## compactness_worst   1      0.0466  1.2592 -2248.9
## fractal_dimension_worst 1      0.0049  1.2175 -2262.3

reg10 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis-
fractal_dimension_se-compactness_se, data=train)
drop1(reg10)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
##      concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
##      +
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean -
##      diagnosis - fractal_dimension_se - compactness_se
##              Df Sum of Sq      RSS      AIC
## <none>                                1.2138 -2263.5
## perimeter_mean      1      9.7624 10.9762 -1389.1
## area_mean           1      0.0783  1.2921 -2240.6

```

```
## smoothness_mean      1      0.0286  1.2424 -2256.2
## compactness_mean     1      0.5372  1.7510 -2119.7
## concavity_mean       1      0.2918  1.5056 -2179.8
## symmetry_mean        1      0.0054  1.2192 -2263.7
## fractal_dimension_mean 1      0.0071  1.2209 -2263.2
## radius_se            1      0.0214  1.2352 -2258.5
## texture_se           1      0.0220  1.2358 -2258.3
## perimeter_se         1      0.0504  1.2642 -2249.3
## concavity_se         1      0.0405  1.2543 -2252.4
## concave.points_se    1      0.0530  1.2667 -2248.5
## symmetry_se          1      0.0249  1.2387 -2257.4
## radius_worst         1      0.2670  1.4808 -2186.4
## texture_worst        1      0.0349  1.2486 -2254.2
## perimeter_worst      1      0.0563  1.2700 -2247.5
## area_worst           1      0.1315  1.3452 -2224.6
## smoothness_worst     1      0.0190  1.2327 -2259.3
## compactness_worst    1      0.0464  1.2602 -2250.6
## fractal_dimension_worst 1      0.0052  1.2190 -2263.8
```

```
reg11 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis-
fractal_dimension_se-compactness_se-fractal_dimension_worst, data=train)
drop1(reg11)
```

```
## Single term deletions
```

```
##
```

```
## Model:
```

```
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean -
##      diagnosis - fractal_dimension_se - compactness_se -
fractal_dimension_worst
```

```
##              Df Sum of Sq      RSS      AIC
## <none>                1.2190 -2263.8
## perimeter_mean      1      9.7581 10.9771 -1391.1
## area_mean           1      0.0795  1.2985 -2240.7
## smoothness_mean     1      0.0320  1.2510 -2255.5
## compactness_mean    1      0.5852  1.8042 -2109.8
## concavity_mean      1      0.3039  1.5229 -2177.2
## symmetry_mean       1      0.0048  1.2238 -2264.2
## fractal_dimension_mean 1      0.0021  1.2211 -2265.1
```

```
## radius_se          1    0.0211  1.2401 -2259.0
## texture_se         1    0.0232  1.2422 -2258.3
## perimeter_se       1    0.0499  1.2689 -2249.8
## concavity_se        1    0.0422  1.2612 -2252.2
## concave.points_se   1    0.0532  1.2722 -2248.8
## symmetry_se         1    0.0218  1.2408 -2258.7
## radius_worst        1    0.2626  1.4816 -2188.2
## texture_worst       1    0.0352  1.2542 -2254.4
## perimeter_worst     1    0.0542  1.2732 -2248.5
## area_worst          1    0.1298  1.3488 -2225.5
## smoothness_worst    1    0.0220  1.2410 -2258.7
## compactness_worst   1    0.0526  1.2716 -2249.0

reg12 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis-
fractal_dimension_se-compactness_se-fractal_dimension_worst-
fractal_dimension_mean, data=train)
drop1(reg12)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean -
##      diagnosis - fractal_dimension_se - compactness_se -
fractal_dimension_worst -
##      fractal_dimension_mean
##              Df Sum of Sq      RSS      AIC
## <none>                    1.2211 -2265.1
## perimeter_mean      1    10.4460 11.6672 -1368.8
## area_mean           1     0.0878  1.3090 -2239.4
## smoothness_mean     1     0.0343  1.2554 -2256.1
## compactness_mean    1     0.8296  2.0507 -2060.8
## concavity_mean      1     0.3054  1.5265 -2178.3
## symmetry_mean       1     0.0047  1.2259 -2265.6
## radius_se           1     0.0203  1.2414 -2260.6
## texture_se          1     0.0249  1.2460 -2259.1
## perimeter_se        1     0.0491  1.2703 -2251.4
## concavity_se        1     0.0466  1.2677 -2252.2
## concave.points_se   1     0.0518  1.2729 -2250.6
```

```

## symmetry_se      1      0.0221  1.2432 -2260.0
## radius_worst     1      0.2719  1.4930 -2187.1
## texture_worst    1      0.0384  1.2596 -2254.8
## perimeter_worst  1      0.0568  1.2780 -2249.0
## area_worst       1      0.1315  1.3526 -2226.4
## smoothness_worst 1      0.0214  1.2425 -2260.2
## compactness_worst 1      0.0510  1.2721 -2250.8

reg13 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis-
fractal_dimension_se-compactness_se-fractal_dimension_worst-
fractal_dimension_mean-symmetry_mean, data=train)
drop1(reg13)

## Single term deletions
##
## Model:
## radius_mean ~ (diagnosis + texture_mean + perimeter_mean + area_mean +
##      smoothness_mean + compactness_mean + concavity_mean +
concave.points_mean +
##      symmetry_mean + fractal_dimension_mean + radius_se + texture_se +
##      perimeter_se + area_se + smoothness_se + compactness_se +
##      concavity_se + concave.points_se + symmetry_se + fractal_dimension_se
+
##      radius_worst + texture_worst + perimeter_worst + area_worst +
##      smoothness_worst + compactness_worst + concavity_worst +
##      concave.points_worst + symmetry_worst + fractal_dimension_worst) -
##      texture_mean - concave.points_worst - area_se - smoothness_se -
##      symmetry_worst - concavity_worst - concave.points_mean -
##      diagnosis - fractal_dimension_se - compactness_se -
fractal_dimension_worst -
##      fractal_dimension_mean - symmetry_mean
##              Df Sum of Sq      RSS      AIC
## <none>                                1.2259 -2265.6
## perimeter_mean      1    10.4955 11.7214 -1369.0
## area_mean           1     0.0908  1.3167 -2239.1
## smoothness_mean     1     0.0405  1.2663 -2254.6
## compactness_mean    1     0.8249  2.0508 -2062.8
## concavity_mean      1     0.3006  1.5265 -2180.3
## radius_se           1     0.0241  1.2499 -2259.8
## texture_se          1     0.0239  1.2497 -2259.9
## perimeter_se        1     0.0544  1.2803 -2250.3
## concavity_se        1     0.0442  1.2700 -2253.5
## concave.points_se   1     0.0514  1.2773 -2251.2
## symmetry_se         1     0.0175  1.2434 -2261.9
## radius_worst        1     0.2729  1.4987 -2187.6
## texture_worst       1     0.0376  1.2634 -2255.5
## perimeter_worst     1     0.0549  1.2808 -2250.1
## area_worst          1     0.1370  1.3628 -2225.4

```

```
## smoothness_worst 1 0.0232 1.2490 -2260.1
## compactness_worst 1 0.0529 1.2788 -2250.7
```

with reg13 me, rmse and mape were slightly reduced compared to the original dataset

```
#####
## 5. testing the quality of both regression models (The original and the latest one)
#####
```

```
predict = predict(reg13, eval)
me = mean(eval$radius_mean-predict)
me
```

```
## [1] 0.0102765
```

```
rmse = sqrt(mean((eval$radius_mean-predict)**2))
rmse
```

```
## [1] 0.06643954
```

```
mape = mean(abs(eval$radius_mean-predict)/eval$radius_mean)*100
mape
```

```
## [1] 0.3295522
```

```
predictCV = length(df)
```

```
#####
## 6. Perform cross validation on both models
#####
```

##cross-validation (full) - Leave-one-out-method

```
for (k in 1:length(df$radius_mean)) {
  eval = df[k,]
  train = df[-k,]
  # here we use the Leave-one-out-method on the whole dataFrame
  m1 = lm(radius_mean~.,data=train)
  predictCV[k] = predict(m1,eval)
}
```

```
me = mean(df$radius_mean-predictCV)
me
```

```
## [1] -0.00007908341
```

```
rmse = sqrt(mean((df$radius_mean-predictCV)**2))
rmse
```

```
## [1] 0.06459013
```



```

mape = mean(abs(df$radius_mean-predictCV)/df$radius_mean)*100
mape

## [1] 0.3030293

##cross-validation(reduced)
for (k in 1:length(df$radius_mean)) {
  eval = df[k,]
  train = df[-k,]
# here we exclude the columns we identified as not necessary
  m1 = lm(radius_mean~.-texture_mean-concave.points_worst-area_se-
smoothness_se-symmetry_worst-concavity_worst-concave.points_mean-diagnosis-
fractal_dimension_se-compactness_se-fractal_dimension_worst-
fractal_dimension_mean-symmetry_mean, data=train)
  predictCV[k] = predict(m1,eval)
}

me = mean(df$radius_mean-predictCV)
me

## [1] 0.0001016514

rmse = sqrt(mean((df$radius_mean-predictCV)**2))
rmse

## [1] 0.06221155

mape = mean(abs(df$radius_mean-predictCV)/df$radius_mean)*100
mape

## [1] 0.2961335

```