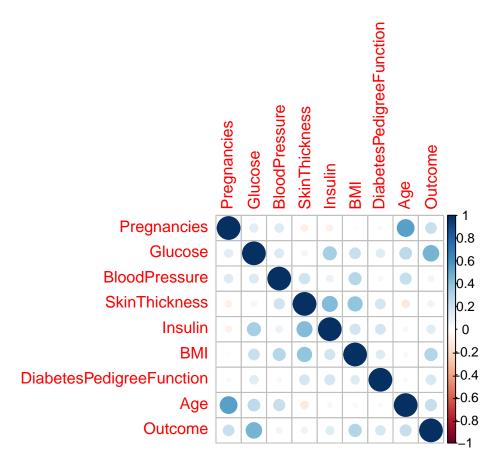
## Analysis using Logistic Regression

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## ## [1] FALSE

##	Pregnancies	Glucose	BP	ST	
##	Min. : 0.000	Min. : 0.0	Min. : 0.00	Min. : 0.00	
##	1st Qu.: 1.000	1st Qu.: 99.0	1st Qu.: 62.00	1st Qu.: 0.00	
##	Median : 3.000	Median :117.0	Median : 72.00	Median :23.00	
##	Mean : 3.845	Mean :120.9	Mean : 69.11	Mean :20.54	
##	3rd Qu.: 6.000	3rd Qu.:140.2	3rd Qu.: 80.00	3rd Qu.:32.00	
##	Max. :17.000	Max. :199.0	Max. :122.00	Max. :99.00	
##	Insulin	BMI	DPF	Age	Outcome
##	Min. : 0.0	Min. : 0.00	Min. :0.0780	Min. :21.00	0:500
##	1st Qu.: 0.0	1st Qu.:27.30	1st Qu.:0.2437	1st Qu.:24.00	1:268
##	Median : 30.5	Median :32.00	Median :0.3725	Median :29.00	
##	Mean : 79.8	Mean :31.99	Mean :0.4719	Mean :33.24	
##	3rd Qu.:127.2	3rd Qu.:36.60	3rd Qu.:0.6262	3rd Qu.:41.00	
##	Max. :846.0	Max. :67.10	Max. :2.4200	Max. :81.00	

Unbalanced distribution, which means about 65% people in this dataset did not have diabetes. Given the Y(outcome) variable is categorical, we would need to use the logistic regression model. Using undersampling to reduce bias towards the majority.

```
## Training
##
##
     0
         1
## 215 215
## Testing
##
##
   0
       1
## 53 53
## [1] "train sample size:
                            430"
  [1] "test sample size:
                           106"
##
##
    0
         1
## 215 215
##
##
   0
       1
## 53 53
Generalized Linear Model
Logistic Regression
Using Logit:
##
## Call:
  glm(formula = Outcome ~ Pregnancies + Glucose + BP + Insulin +
##
       BMI + DPF + Age, family = binomial, data = diabetes.training)
##
## Deviance Residuals:
       Min
##
                   1Q
                         Median
                                        30
                                                 Max
##
  -2.95529 -0.77910 -0.00446
                                   0.74787
                                             2.71693
##
## Coefficients:
##
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) -8.347946
                           0.952058
                                     -8.768
                                              < 2e-16 ***
## Pregnancies 0.106247
                           0.042935
                                       2.475
                                             0.01334 *
## Glucose
                0.035361
                           0.004826
                                      7.327 2.36e-13 ***
## BP
               -0.014462
                           0.006700
                                     -2.159
                                             0.03088 *
## Insulin
               -0.001951
                           0.001074
                                      -1.817 0.06915 .
## BMI
                0.090030
                           0.019238
                                       4.680 2.87e-06 ***
## DPF
                1.262500
                           0.401056
                                       3.148 0.00164 **
                                       2.441 0.01465 *
## Age
                0.031893
                           0.013066
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
  (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 596.11 on 429 degrees of freedom
```

```
## Residual deviance: 424.23 on 422 degrees of freedom
## AIC: 440.23
##
## Number of Fisher Scoring iterations: 5
```

The StepAIC function was used to determine the model goodness of fit between the logit and probit model. This decided the outcome that the logit model is suitable for this specific task as it has a lower AIC compared to the probit model. Also the insignificant variables are the skin thickness and age.

The logistic regression coefficients give the change in the log odds of the outcome for a one unit increase in the predictor variable.

The maximum likelihood estimation can be expressed as:

```
\ln\frac{\pi}{1-\pi} = -8.653 + 0.113X_1 + 0.0438X_2 - 0.0110X_3 - 0.00224X_4 + 0.0938X_5 + 1.174X_6 ## (Intercept) Pregnancies Glucose BP Insulin BMI ## 0.0002368826 1.1120965563 1.0359936362 0.9856419140 0.9980507767 1.0942070495 BF DPF Age ## 3.5342470033 1.0324074648
```

Interpretation of step model:

- For every one unit increase in pregnancies, there is an increase change in (1.12 1) \* 100 = 12% in odds ratio
- For every one unit increase in glucose, there is an increase change in (1.04-1)\*100 = 4% in odds ratio
- For every one unit increase in BP, there is decrease change of 1.1% in odds ratio
- For every one unit increase in Insulin, there is decrease change of 0.2% in odds ratio
- For every one unit increase in BMI, there is an increase change of 9.8% in odds ratio
- For every one unit increase in DPF, there is an increase change in 223% in odds ratio

## Statistical Inference:

```
##
                       2.5 %
                                     97.5 %
## (Intercept) -10.304333456 -6.5643927482
                 0.023037125
                              0.1917972781
## Pregnancies
## Glucose
                 0.026289814
                              0.0452457449
## BP
                -0.028127001 -0.0017158519
## Insulin
                -0.004048190
                              0.0001865544
## BMI
                 0.053878350
                              0.1294103525
## DPF
                 0.489182126
                              2.0637607030
## Age
                 0.006631532 0.0580279643
```

## Prediction:

```
## pred.classes
## Pred. neg Pred. pos
## Obs. neg 39 14
## Obs. pos 17 36
## [1] 0.7075472
```