

Final_Report

2019

Importing data into R

```
setwd ("/Users/khushbu_desai/Documents/KHUSHBU/MS/py_scripts_main/GitHub/Predicting-Newborn-Babies-Birthweight/hs631_final_proj_report.html  
to_Notebook_Python_Scripting")
```

```
library(ggplot2)  
library(rgl)  
library(corrplot)
```

```
## corrplot 0.84 loaded
```

```
library(pwr)  
library(pscl)
```

```
## Classes and Methods for R developed in the  
## Political Science Computational Laboratory  
## Department of Political Science  
## Stanford University  
## Simon Jackman  
## hurdle and zeroinfl functions by Achim Zeileis
```

```
library(car)
```

```
## Loading required package: carData
```

```

logisticPseudoR2s <- function(LogModel) {
  dev <- LogModel$deviance
  nullDev <- LogModel$null.deviance
  modelN <- length(LogModel$fitted.values)
  R.l <- 1 - dev / nullDev
  R.cs <- 1- exp ( -(nullDev - dev) / modelN)
  R.n <- R.cs / ( 1 - ( exp (-(nullDev / modelN))))
  cat("Pseudo R^2 for logistic regression\n")
  cat("Hosmer and Lemeshow R^2   ", round(R.l, 3), "\n")
  cat("Cox and Snell R^2         ", round(R.cs, 3), "\n")
  cat("Nagelkerke R^2           ", round(R.n, 3), "\n")
}

```

```

load("births.rdat")
summary(births)

```

```

##      dad_age      mom_age      maturity      len_preg
##  Min.   : 14.0    Min.    :13    advnced:133    Min.    : 20.00
##  1st Qu.: 26.0    1st Qu.:22    younger:867  1st Qu.: 37.00
##  Median : 32.0    Median  :27                      Median : 39.00
##  Mean   :195.9    Mean    :27                      Mean   : 40.26
##  3rd Qu.: 39.0    3rd Qu.:32                      3rd Qu.: 40.00
##  Max.   :999.0    Max.    :50                      Max.   :999.00
##      is_premie      num_visits      marital      mom_wt_gain
##  fullterm:846    Min.    : 0.00    married :613    Min.    : 0.00
##  premie  :152    1st Qu.: 10.00    unknown : 1    1st Qu.: 21.00
##  unknown : 2    Median : 12.00    unmarried:386  Median : 30.00
##                      Mean   : 20.99                      Mean   : 56.48
##                      3rd Qu.: 15.00                      3rd Qu.: 40.00
##                      Max.   :999.00                      Max.   :999.00
##      bwt      low_bwt      sex      smoke      mom_white
##  Min.    : 1.000    low    :111    female:503    nonsmoker:873    nonwhite:284
##  1st Qu.: 6.380    notlow:889    male  :497    smoker  :126    unknown : 2
##  Median : 7.310                      unknown : 1    white   :714
##  Mean    : 7.101
##  3rd Qu.: 8.060
##  Max.    :11.750
##  mom_age_level
##  35+      :133
##  early20s:281
##  early30s:219
##  late20s :257
##  teens   :110
##

```

Data Cleaning

```
# Cleaning impossible values
births$dad_age[births$dad_age == 999] <- NA
births$len_preg[births$len_preg == 999] <- NA
births$sis_premie[births$sis_premie == "unknown"] <- NA
births$num_visits[births$num_visits == 999] <- NA
births$marital[births$marital == "unknown"] <- NA
births$mom_wt_gain[births$mom_wt_gain == 999] <- NA
births$smoke[births$smoke == "unknown"] <- NA
births$mom_white[births$mom_white == "unknown"] <- NA

# Drop unused levels
births$sis_premie <- droplevels(births$sis_premie)
births$marital <- droplevels(births$marital)
births$smoke <- droplevels(births$smoke)
births$mom_white <- droplevels(births$mom_white)

# Reordering levels for correct visualizations

births$maturity <- factor(births$maturity, levels = c("younger","advnced"))
births$mom_age_level <- factor( births$mom_age_level, levels = c("teens","early20s","late20s","early30s","35+") )
births$sis_premie <- factor(births$sis_premie, levels = c("premie","fullterm"))
```

EDA

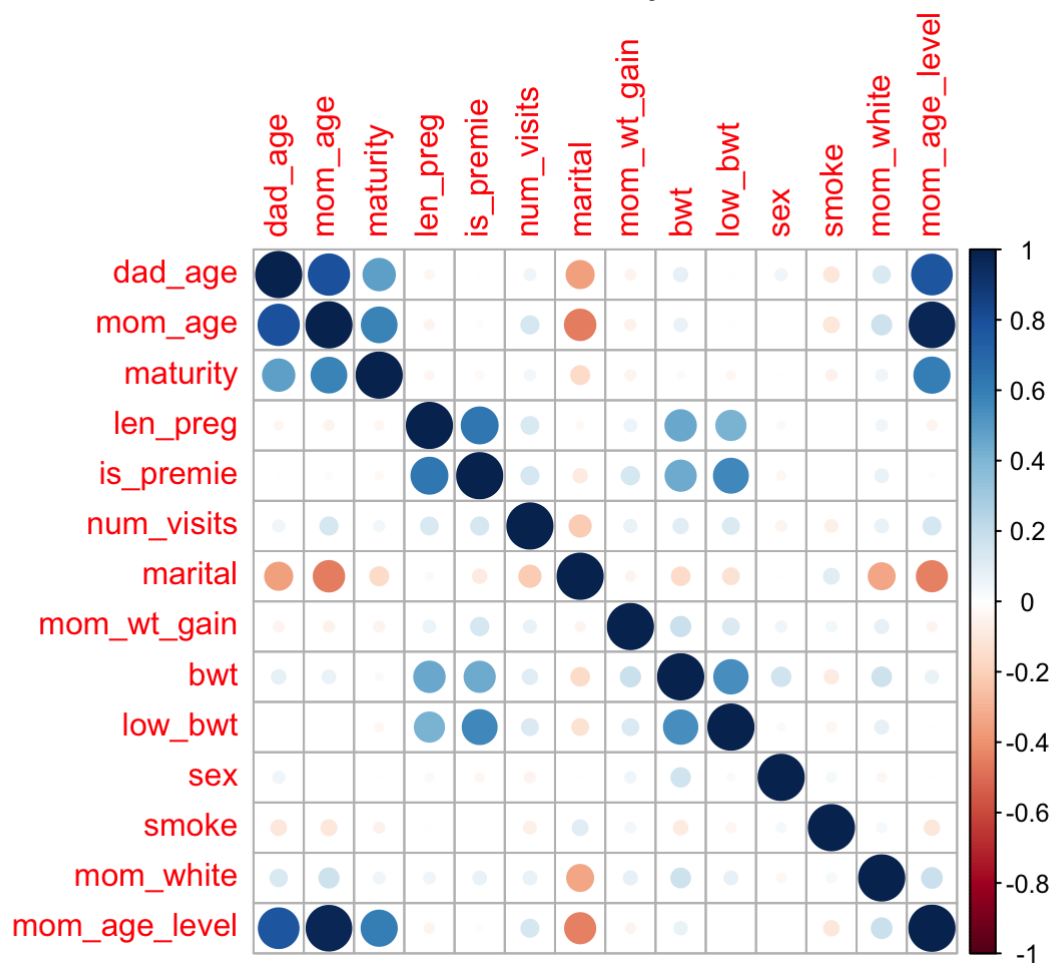
```
# Summary Statistics : mean, median & quantiles included in summary.for categorical variable counts are included too.
summary(births)
```

```
##      dad_age      mom_age      maturity      len_preg      is_premie
## Min.      :14.00    Min.      :13    younger:867    Min.      :20.00    premie   :152
## 1st Qu.:25.00    1st Qu.:22    advncd:133    1st Qu.:37.00    fullterm:846
## Median :30.00    Median :27                                Median :39.00    NA's      : 2
## Mean   :30.26    Mean   :27                                Mean   :38.33
## 3rd Qu.:35.00    3rd Qu.:32                                3rd Qu.:40.00
## Max.   :55.00    Max.   :50                                Max.   :45.00
## NA's   :171                                NA's   :2
##      num_visits      marital      mom_wt_gain      bwt
## Min.      : 0.0    married :613    Min.      : 0.00    Min.      : 1.000
## 1st Qu.:10.0    unmarried:386    1st Qu.:20.00    1st Qu.: 6.380
## Median :12.0    NA's      : 1    Median :30.00    Median : 7.310
## Mean   :12.1                                Mean   :30.33    Mean   : 7.101
## 3rd Qu.:15.0                                3rd Qu.:38.00    3rd Qu.: 8.060
## Max.   :30.0                                Max.   :85.00    Max.   :11.750
## NA's   :9                                NA's   :27
##      low_bwt      sex      smoke      mom_white      mom_age_level
## low      :111    female:503    nonsmoker:873    nonwhite:284    teens      :110
## notlow:889    male  :497    smoker      :126    white      :714    early20s:281
##                                     NA's      : 1    NA's      : 2    late20s :257
##                                     NA's      : 2    early30s:219
##                                     35+      :133
##
##
```

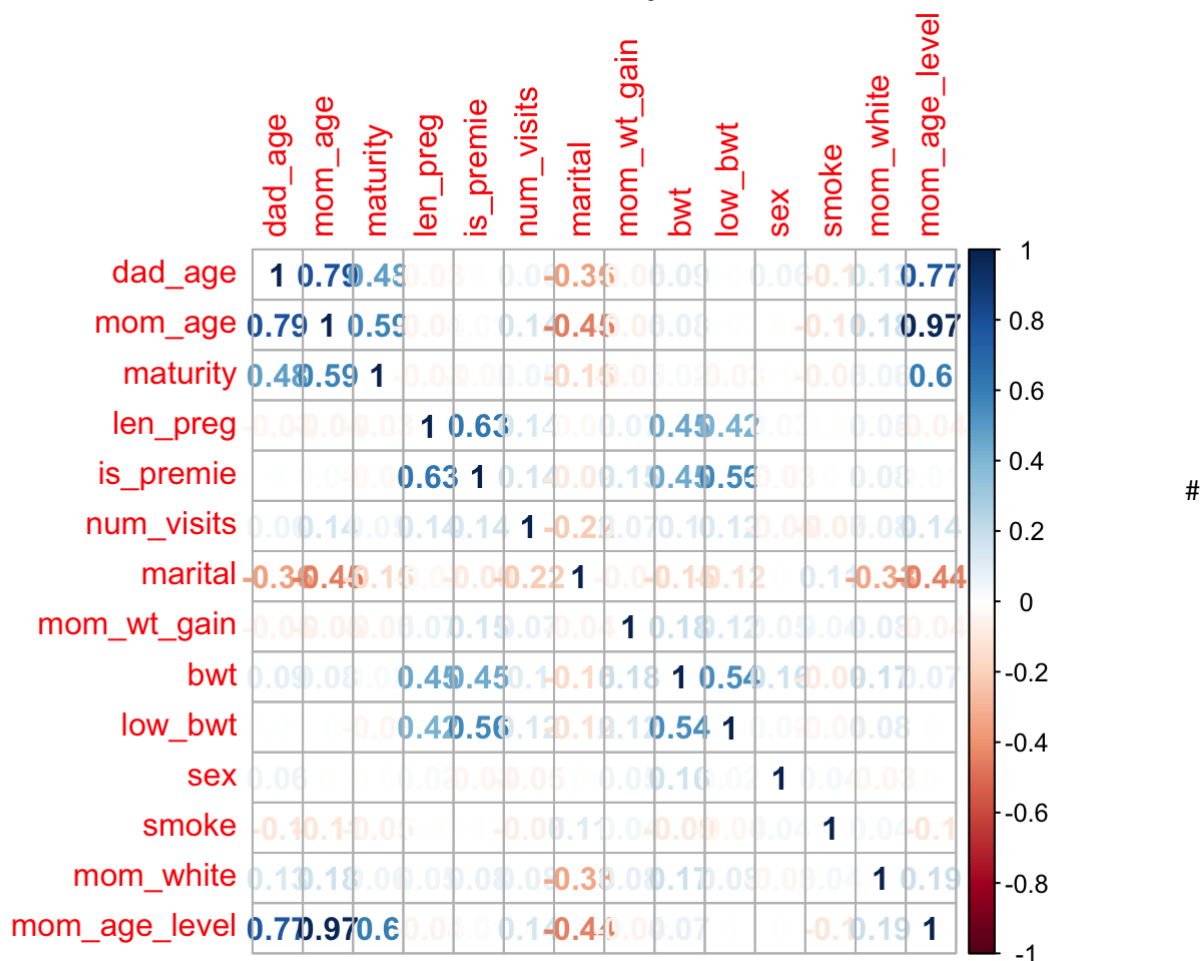
Correlation

```
births_numeric <- births[,c(1:14)]
births_numeric$maturity <- as.numeric(births$maturity )
births_numeric$sis_premie <- as.numeric(births$sis_premie )
births_numeric$marital <- as.numeric(births$marital )
births_numeric$low_bwt <- as.numeric(births$low_bwt )
births_numeric$ssex <- as.numeric(births$ssex )
births_numeric$smoke <- as.numeric(births$smoke )
births_numeric$mom_white <- as.numeric(births$mom_white)
births_numeric$mom_age_level <- as.numeric(births$mom_age_level)

c <- cor(births_numeric, use = "pairwise.complete.obs", method = "spearman")
corrplot(c)
```



```
corrplot(c, method="number")
```

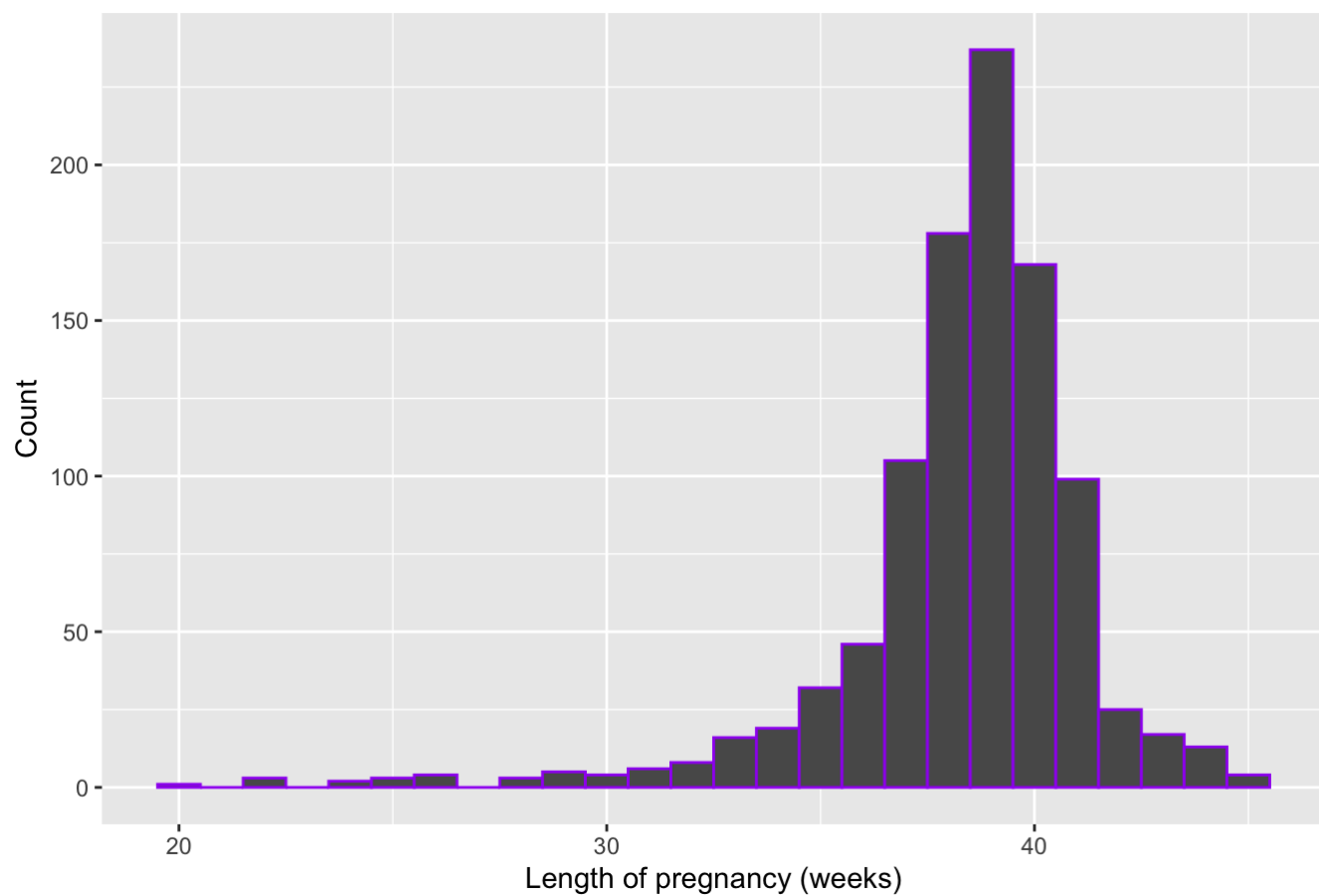


Data Visualizations

```
g4 = ggplot(births, aes(x=len_preg))
g4 + geom_histogram(binwidth = 1,color="purple") + ggtitle("Distribution of the Length of pregnancy") +
  xlab("Length of pregnancy (weeks)") + ylab("Count")
```

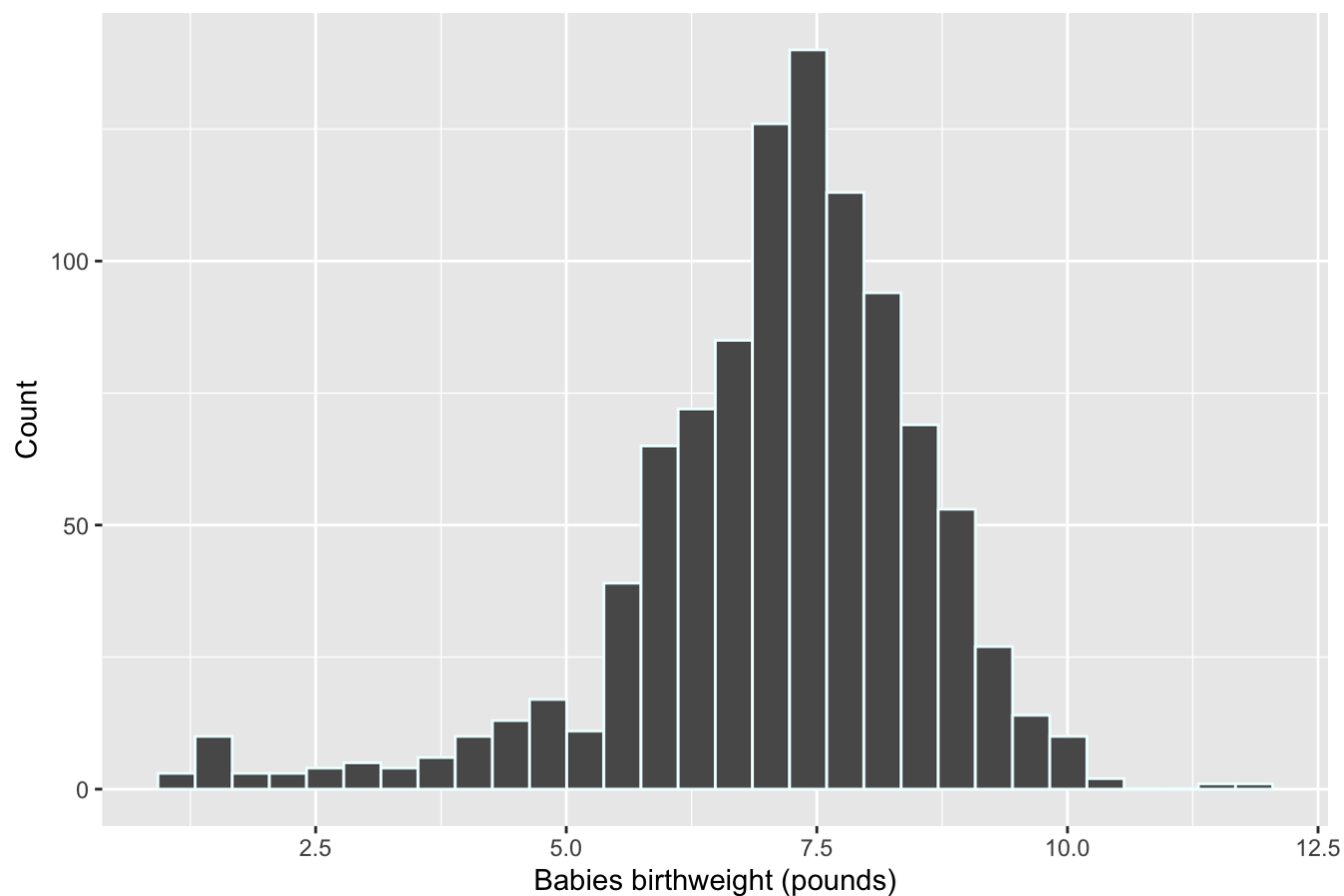
```
## Warning: Removed 2 rows containing non-finite values (stat_bin).
```

Distribution of the Length of pregnancy



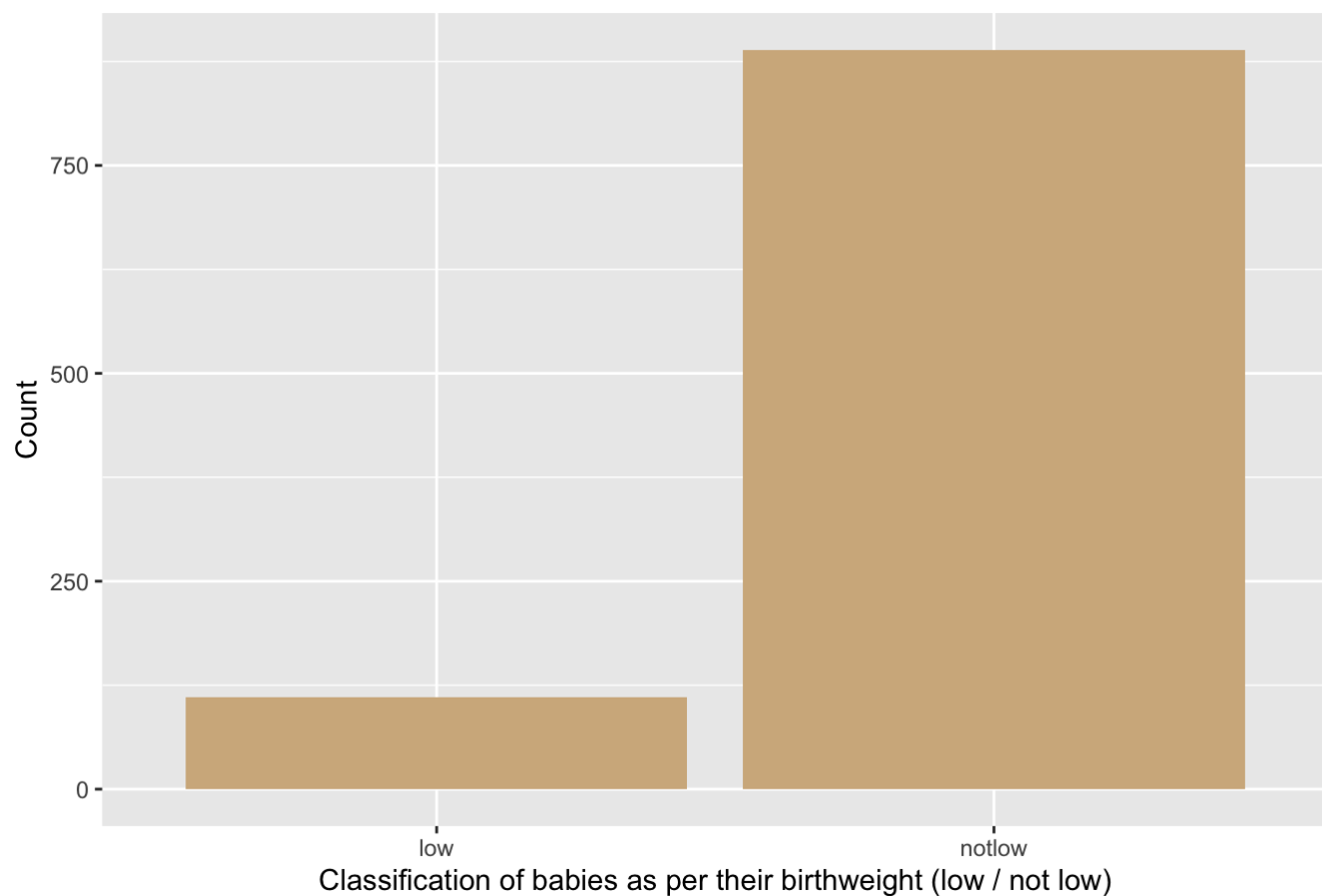
```
g9 = ggplot(births, aes(x=bwt))
g9 + geom_histogram(bins = 30, color = "azure") + ggtitle("Distribution of babies
birthweight") +
  xlab("Babies birthweight (pounds)") + ylab("Count")
```

Distribution of babies birthweight



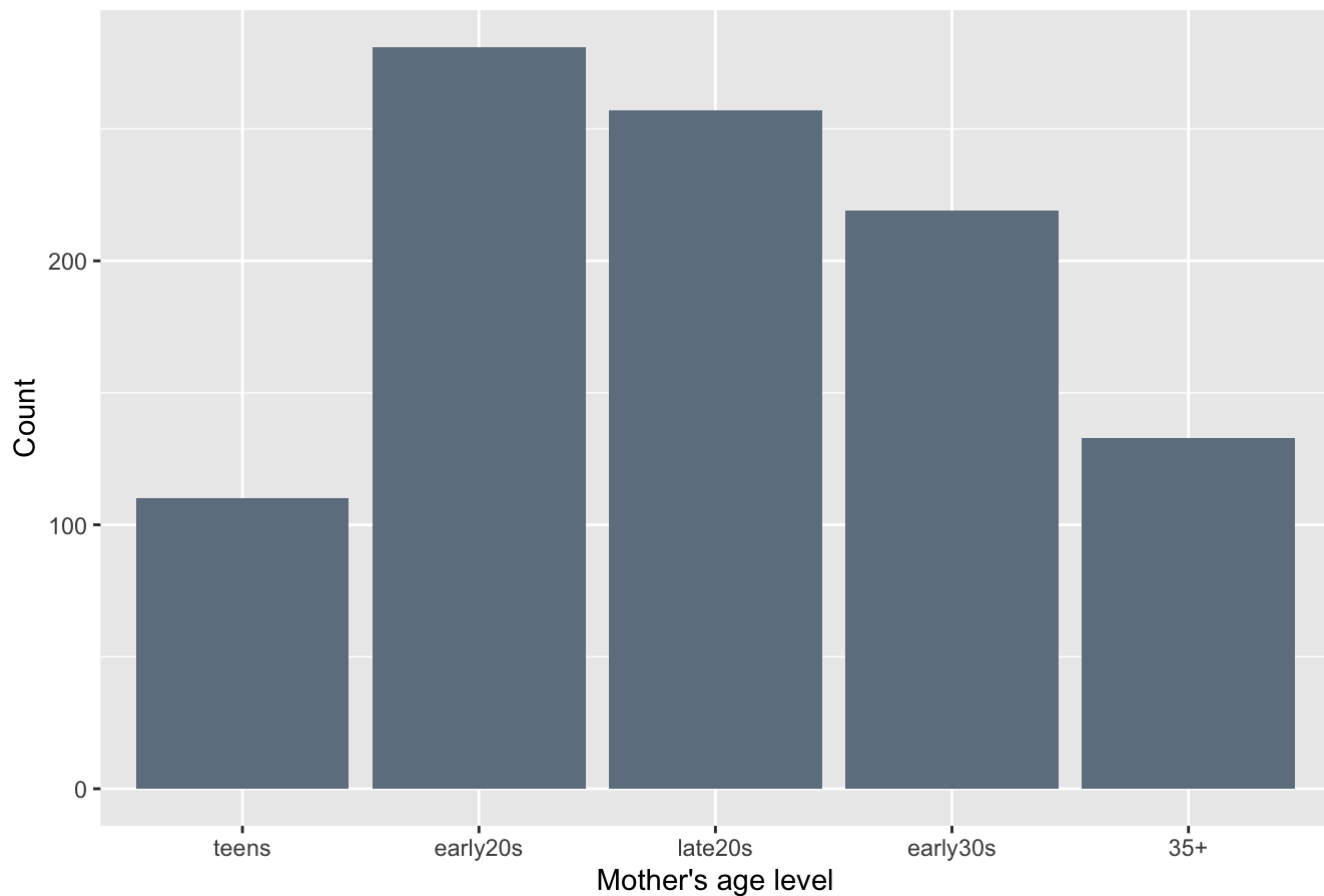
```
g10 = ggplot(births, aes(x=low_bwt))
g10 + geom_bar(fill = "tan")+ ggtitle("Distribution of babies as per their birthweight") +
  xlab("Classification of babies as per their birthweight (low / not low)") + ylab("Count")
```


Distribution of babies as per their birthweight



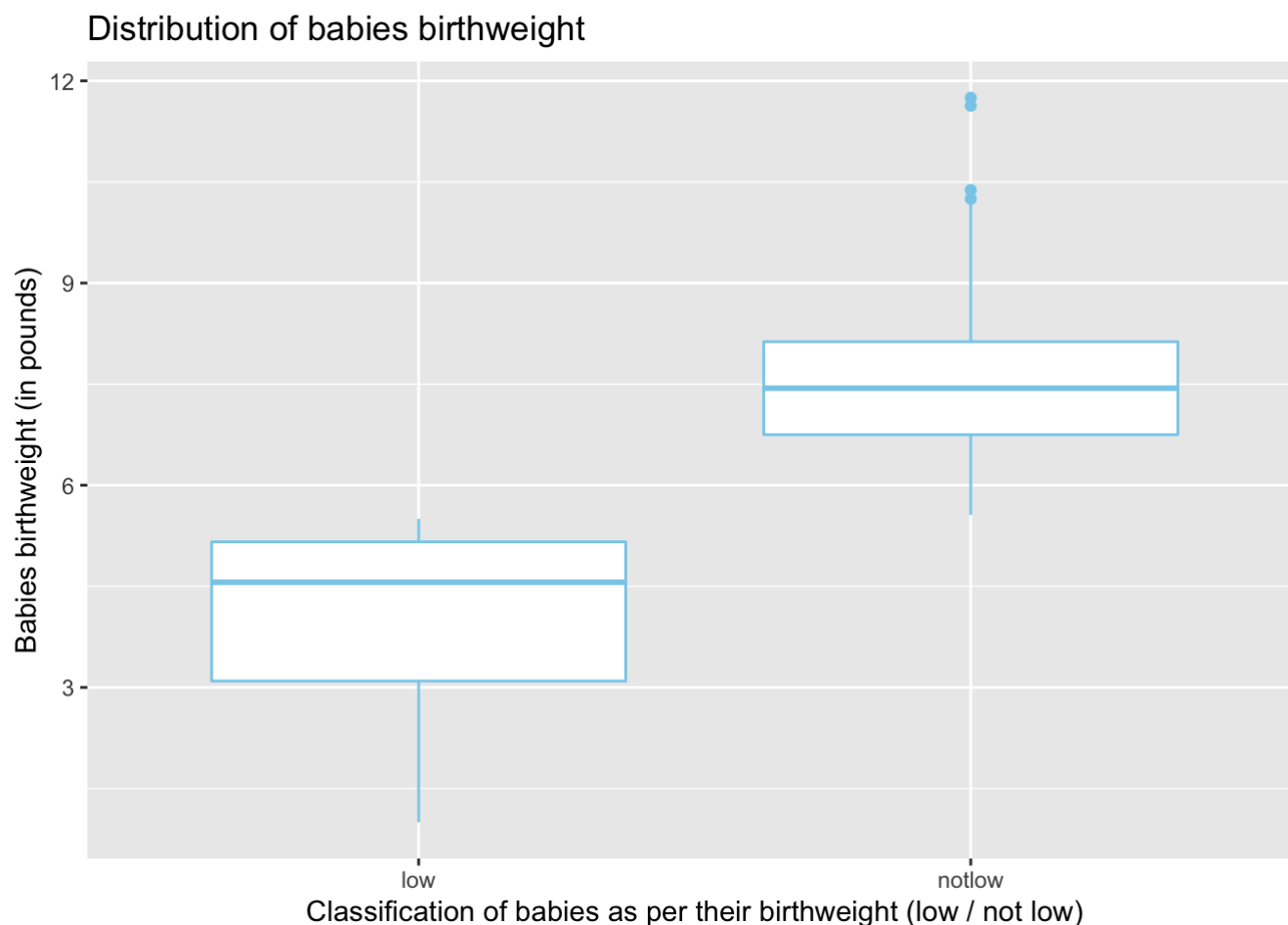
```
g14 = ggplot(births, aes(x=mom_age_level))
g14 + geom_bar(fill = "slategrey")+ ggtitle("Distribution of mother's age level")
+
  xlab("Mother's age level") + ylab("Count")
```

Distribution of mother's age level



the sample has random selectipn of mother's accross all age levels

```
g15 = ggplot(births, aes(x=low_bwt, y=bwt))
g15 + geom_boxplot(color="skyblue")+ ggtitle("Distribution of babies birthweight")
+
  xlab("Classification of babies as per their birthweight (low / not low)") + ylab
("Babies birthweight (in pounds)")
```



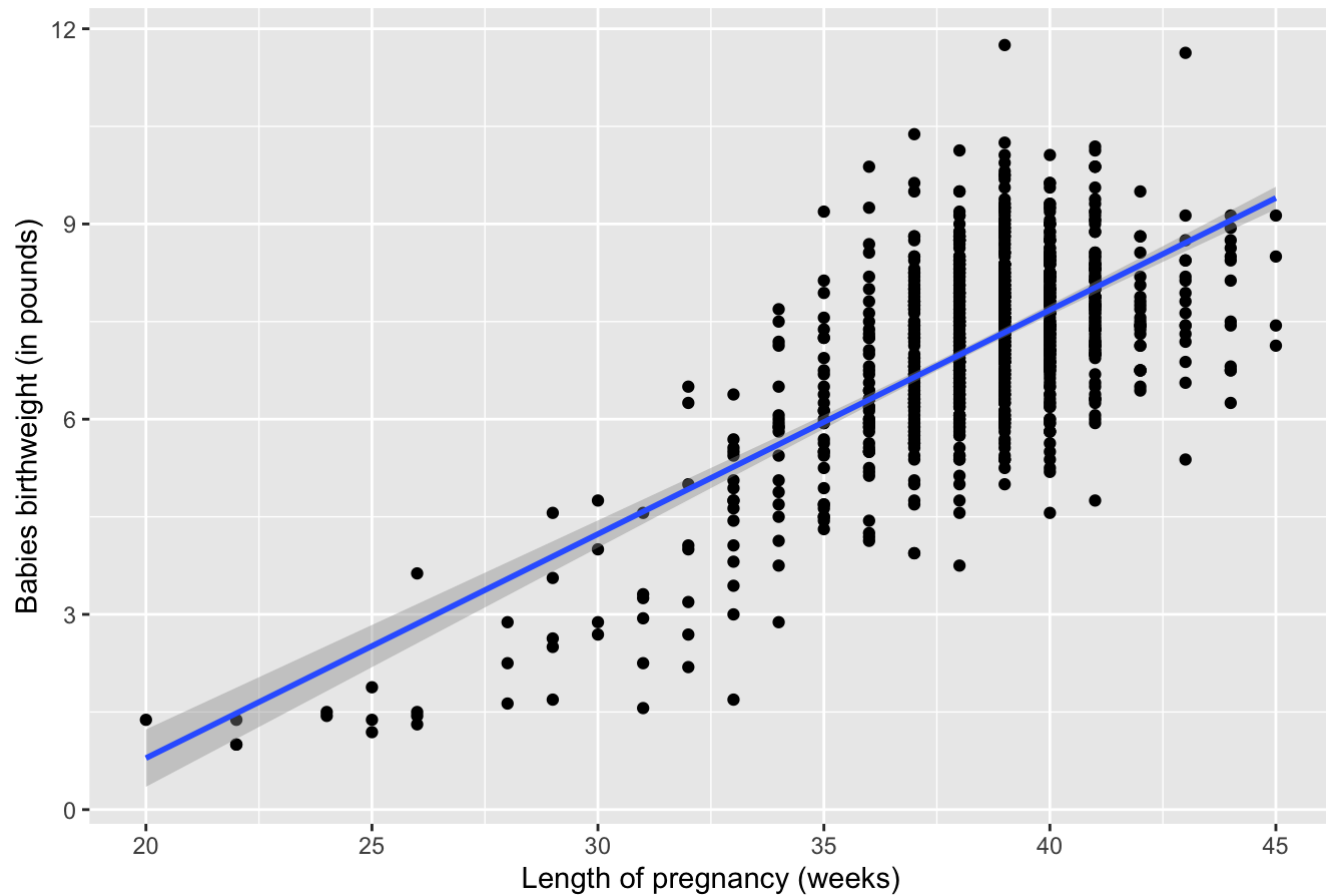
plot implies that the low_bwt variable may be a derived column from the bwt values

```
g16 = ggplot(births, aes(x=len_preg, y= bwt))
g16 + geom_point()+ geom_smooth(method=lm) + ggtitle("Distribution of babies birth
weight and length of pregnancy") +
  xlab("Length of pregnancy (weeks)") + ylab("Babies birthweight (in pounds)")
```

Warning: Removed 2 rows containing non-finite values (stat_smooth).

Warning: Removed 2 rows containing missing values (geom_point).

Distribution of babies birthweight and length of pregnancy

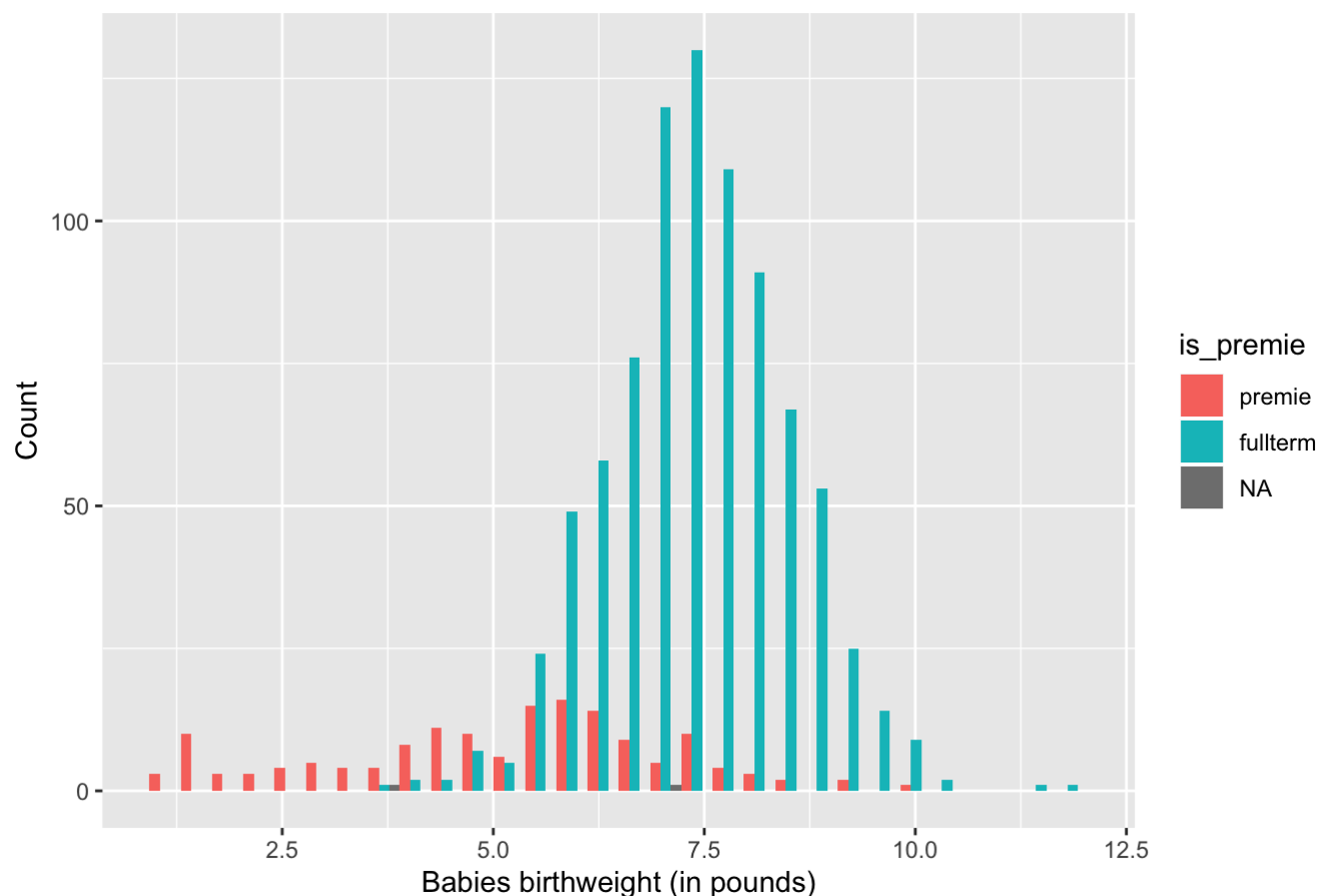


plot implies that there is positive correlation between the length of pregnancy and babies birthweight

the more the length of pregnancy, more will be the babies birthweight

```
g17 = ggplot(births, aes(x=bwt, fill=is_premie))
g17 + geom_histogram(bins=30, position="dodge") + ggtitle("Distribution of babies birthweight and baby being premature or not") +
  xlab("Babies birthweight (in pounds)") + ylab("Count")
```

Distribution of babies birthweight and baby being premature or not

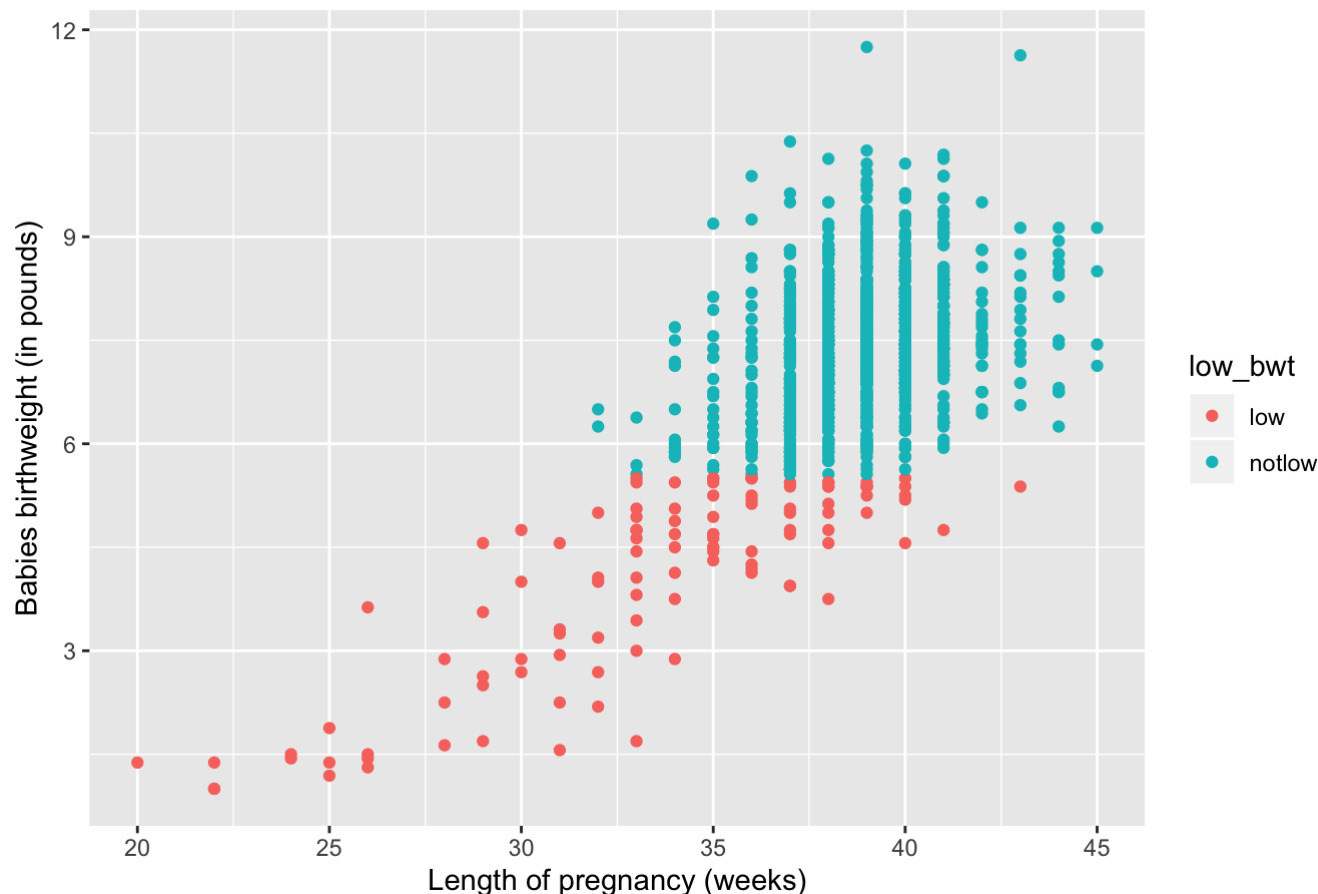


plot implies that there is positive correlation between the babies birthweight and the baby being premature or fullterm
the fullterm babies will have more birthweight, than the premature babies with lower birthweights

```
g18 = ggplot(births, aes(x=len_preg, y= bwt, col=low_bwt))
g18 + geom_point() + ggtitle("Distribution of babies birthweight and length of pregnancy, classified by low birthweight or not") +
  xlab("Length of pregnancy (weeks)") + ylab("Babies birthweight (in pounds)")
```

```
## Warning: Removed 2 rows containing missing values (geom_point).
```

Distribution of babies birthweight and length of pregnancy, classified by low birth

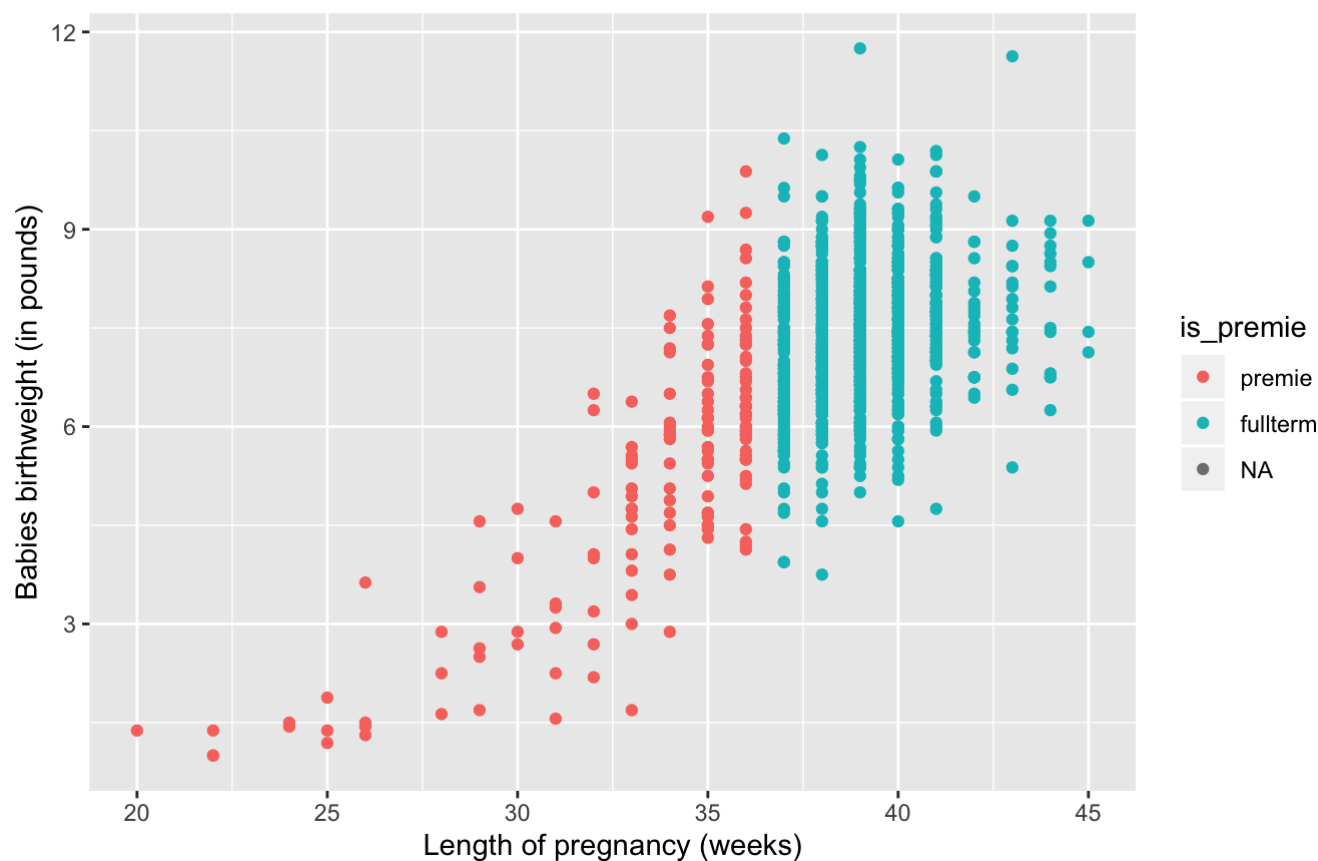


```
# plot implies that all the babies delivered after a longer length of pregnancy, have a higher birthweight
# all the babies delivered after a short length of pregnancy, have a low birthweight
# all babies with low birthweights can be classified as low_bwt babies
```

```
g19 = ggplot(births, aes(x=len_preg, y= bwt, col=is_premie))
g19 + geom_point() + ggtitle("Distribution of babies birthweight and length of pregnancy,
classifying babies as premature or not") + xlab("Length of pregnancy (weeks)") +
ylab("Babies birthweight (in pounds)")
```

```
## Warning: Removed 2 rows containing missing values (geom_point).
```

Distribution of babies birthweight and length of pregnancy, classifying babies as premature or not



```
# plot implies that all the babies delivered after a longer length of pregnancy, have a higher birthweight
# and are fullterm babies
# all babies with shorter length of pregnancy and low birthweights are premature babies
```

Statistical Tests

```
# T-test :
t.test(births$bwt~births$is_premie)
```

```
##
##  Welch Two Sample t-test
##
## data:  births$bwt by births$sis_premie
## t = -14.216, df = 167.51, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -2.654709 -2.007267
## sample estimates:
##      mean in group premie mean in group fullterm
##                5.128421                7.459409
```

```
# conclusions from the t test:
# p-value less than significance level=0.05
# and there is a significant difference in means of both groups : birthweight for
premature and fullterm babies
```

```
table(births$sis_premie)  # n1=152, n2=846
```

```
##
##      premie fullterm
##         152       846
```

```
pwr.t2n.test(n1=152, n2=846, d=0.2, sig.level = .05 )
```

```
##
##      t test power calculation
##
##                n1 = 152
##                n2 = 846
##                d = 0.2
##      sig.level = 0.05
##      power = 0.6210034
##      alternative = two.sided
```

```
# small effect size can be detected and the power of the study is 80%
# indicates results being statistically significant
```

```
# one-way ANOVA:
summary(aov(births$bwt~births$mom_age_level))
```

```
##                Df Sum Sq Mean Sq F value Pr(>F)
## births$mom_age_level    4    14.8    3.699    1.629    0.165
## Residuals              995  2259.6    2.271
```



```
# null hypothesis is that the mean bwt of baby is same for all groups in mom's age levels
# the p-value = 0.165, is more than significance level of 0.05
# we don't have enough evidence to reject the null hypothesis
# concluding that there isn't much impact of mother's age level on the babies birth weight

# Chisq Test:
chisq.test(table(births$low_bwt, births$is_premie))
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: table(births$low_bwt, births$is_premie)
## X-squared = 311.57, df = 1, p-value < 2.2e-16
```

```
# null hypothesis: Both low_bwt and is_premie are independent variables
# the p-value is < 2.2e-16, which is less than the significance level of 0.05
# we have enough evidence to reject the null hypothesis
# and conclude that low_bwt and is_premie are not independent variables
```

Linear Regression model

```
fit_all_lm <- lm(bwt ~ dad_age + mom_age + maturity + len_preg + is_premie + num_v
isits + marital + mom_wt_gain + low_bwt +
                sex + smoke + mom_white + mom_age_level, data = births)
summary(fit_all_lm)
```

```
##
## Call:
## lm(formula = bwt ~ dad_age + mom_age + maturity + len_preg +
##      is_premie + num_visits + marital + mom_wt_gain + low_bwt +
##      sex + smoke + mom_white + mom_age_level, data = births)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.4770 -0.6012 -0.0021  0.5494  3.5563
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -2.6503561   0.7170302   -3.696 0.000234 ***
## dad_age         0.0080555   0.0077365    1.041 0.298087
## mom_age         0.0090994   0.0218822    0.416 0.677644
## maturityadvncd  -0.0731887   0.4196130   -0.174 0.861581
## len_preg        0.1718275   0.0181833    9.450 < 2e-16 ***
## is_premiefullterm 0.0442754   0.1440740    0.307 0.758689
## num_visits      0.0003972   0.0089395    0.044 0.964572
## maritalunmarried -0.1172818   0.0853089   -1.375 0.169590
## mom_wt_gain      0.0071629   0.0023449    3.055 0.002330 **
## low_bwtnotlow    2.3881336   0.1431235   16.686 < 2e-16 ***
## sexmale          0.3821920   0.0653943    5.844 7.45e-09 ***
## smokesmoker      -0.2143158   0.1068650   -2.005 0.045255 *
## mom_whitewhite    0.2458025   0.0807928    3.042 0.002425 **
## mom_age_levelearly20s 0.0825564   0.1538582    0.537 0.591714
## mom_age_levelate20s 0.0187617   0.2309455    0.081 0.935273
## mom_age_levelearly30s -0.0630545   0.3222707   -0.196 0.844929
## mom_age_level35+      NA          NA          NA          NA
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9136 on 784 degrees of freedom
## (200 observations deleted due to missingness)
## Multiple R-squared:  0.6055, Adjusted R-squared:  0.598
## F-statistic: 80.24 on 15 and 784 DF, p-value: < 2.2e-16
```

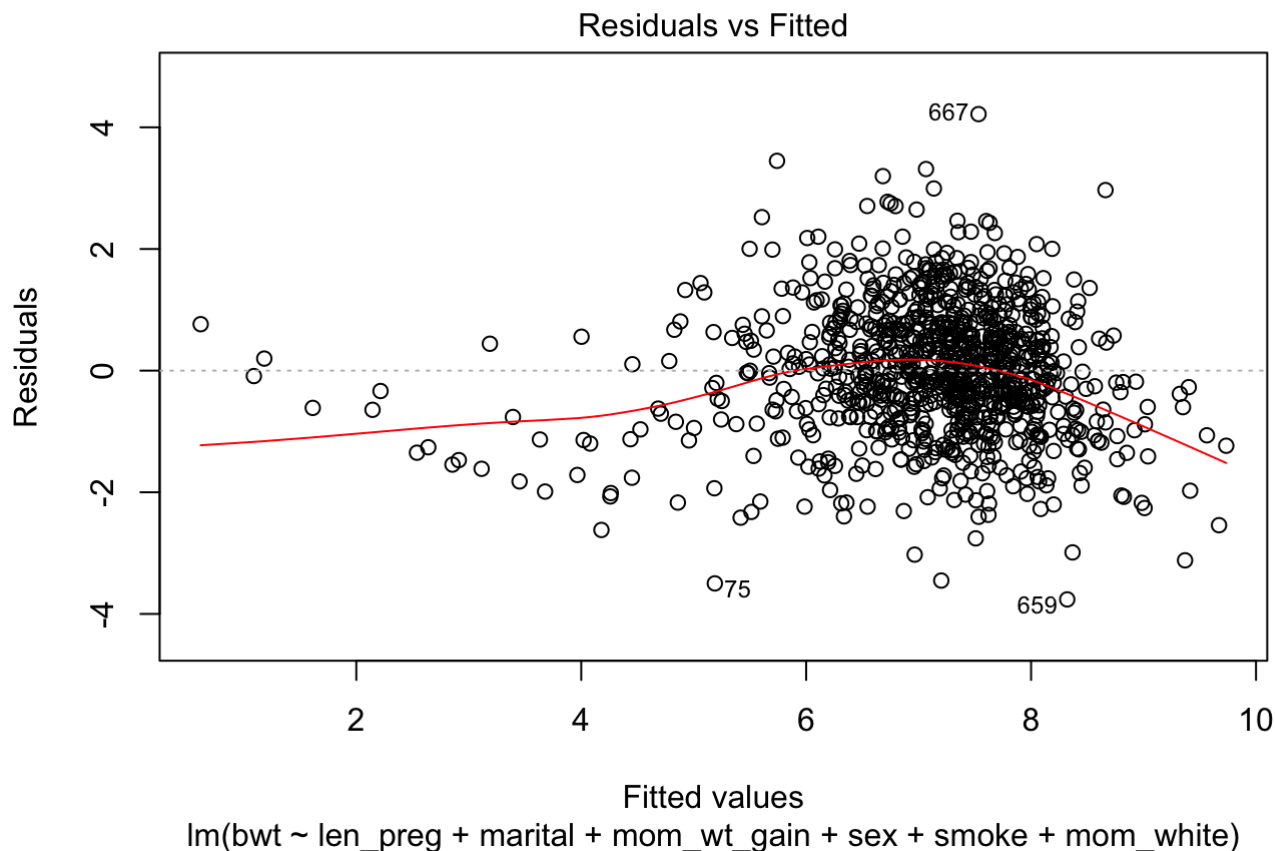
```
# there are variables that have high correlation with other variables, which might
cause multicollinearity
# variables highly correlated (from correlation test above) are :
# maturity with mom_age, is_premie iwth len_preg, low_bwt with bwt and mom_age_lev
els with mom_age
# removing variables that are derived values from other variables: maturity, is_pr
emie, low_bwt, mom_age_levels

# final linear model derived and is statitically significant is as follows:
final_fit_lm <- lm(bwt ~ len_preg + marital + mom_wt_gain + sex + smoke + mom_whit
e, data = births)
summary(final_fit_lm)
```

```
##
## Call:
## lm(formula = bwt ~ len_preg + marital + mom_wt_gain + sex + smoke +
##     mom_white, data = births)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.7613 -0.6582 -0.0263  0.6872  4.2173
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -5.977147   0.462464 -12.925 < 2e-16 ***
## len_preg       0.329042   0.011961  27.509 < 2e-16 ***
## maritalunmarried -0.265136   0.075729  -3.501 0.000485 ***
## mom_wt_gain     0.009258   0.002431   3.809 0.000149 ***
## sexmale        0.378405   0.068971   5.486 5.24e-08 ***
## smokesmoker    -0.388843   0.104758  -3.712 0.000218 ***
## mom_whitewhite  0.212131   0.081324   2.608 0.009236 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.071 on 963 degrees of freedom
## (30 observations deleted due to missingness)
## Multiple R-squared:  0.4869, Adjusted R-squared:  0.4837
## F-statistic: 152.3 on 6 and 963 DF,  p-value: < 2.2e-16
```

```
# all variables have p-value less than 0.05 and their adjusted R squared=0.48
# indicating a good fit model, that will explain 48% variation of bwt (response va
riable) from the total explainable variation,
# when all other variables are held constant

plot(final_fit_lm, 1)
```



```
# the residual values are randomly distributed below & above the fitted line.
# the fit line is close to the regression line
```

```
##### Brief Interpretation of the model #####
```

```
# Birthweight of the baby is dependent on the variables : length of pregnancy(len_preg), marital status of the mother,
# Mother's weight gain before delivery, sex of the baby, mother's smoking status & mother being white or not.
```

```
# linear model: final model's fitted equation:
```

```
# bwt = 0.329042 * len_preg - 0.265136 * marital + 0.009258 * mom_wt_gain + 0.378405 * sex - 0.388843 * smoke + 0.212131 * mom_white - 5.977147
```

```
## For every 1 week increase in the length of pregnancy, the birthweight of the baby increases by 0.329042 pounds on average, when all other variables in the model are held constant
```

```
## For married mothers as compared with unmarried mother's, the birthweight of the baby decreases by 0.265136 pounds on average, when all other variables in the model are held constant
```

Logistic Regression model

```
fit_glm_all <- glm(low_bwt ~ ., family = binomial(), data = births)
```

```
## Warning: glm.fit: algorithm did not converge
```

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

```
summary(fit_glm_all)
```

```
##
## Call:
## glm(formula = low_bwt ~ ., family = binomial(), data = births)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.778e-04  2.100e-08  2.100e-08  2.100e-08  2.431e-04
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -9.264e+02  1.375e+05  -0.007    0.995
## dad_age        -3.359e+00  9.207e+02  -0.004    0.997
## mom_age        -3.718e+00  2.668e+03  -0.001    0.999
## maturityadvncd  1.473e+02  6.438e+04   0.002    0.998
## len_preg       3.282e+00  2.690e+03   0.001    0.999
## is_premiefullterm -2.840e+01  1.373e+04  -0.002    0.998
## num_visits     -1.576e+00  1.079e+03  -0.001    0.999
## maritalunmarried -1.859e+01  6.864e+03  -0.003    0.998
## mom_wt_gain     -5.582e-02  2.668e+02   0.000    1.000
## bwt            1.758e+02  1.299e+04   0.014    0.989
## sexmale        5.662e+00  5.273e+03   0.001    0.999
## smokesmoker    -6.726e+00  6.180e+03  -0.001    0.999
## mom_whitewhite -1.349e+01  7.363e+03  -0.002    0.999
## mom_age_levelearly20s 3.415e+01  1.116e+04   0.003    0.998
## mom_age_levellate20s 9.080e+01  2.875e+04   0.003    0.997
## mom_age_levelearly30s 1.363e+02  4.292e+04   0.003    0.997
## mom_age_level35+           NA           NA           NA           NA
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 4.9325e+02  on 799  degrees of freedom
## Residual deviance: 4.5427e-07  on 784  degrees of freedom
## (200 observations deleted due to missingness)
## AIC: 32
##
## Number of Fisher Scoring iterations: 25
```

```
# summary shows that there are variables in related to the model indicating a perfect linear separation
# the fitted coefficients are not statistically significant to explain the model.
pR2(fit_glm_all)
```

```
##           llh           llhNull           G2           McFadden           r2ML
## -2.271326e-07 -3.486010e+02  6.972020e+02  1.000000e+00  5.816774e-01
##           r2CU
##  1.000000e+00
```

```
# Both McFadden & r2CU values are 1, indicating a model with perfect fit
# which is not possible
```

```
# there are variables that have high correlation with other variables, which might cause multicollinearity
# variables highly correlated (from correlation test above) are :
# maturity with mom_age, is_premie iwth len_preg, low_bwt with bwt and mom_age_levels with mom_age
# removing variables that are derived values from other variables: maturity, is_premie, low_bwt, mom_age_levels
# we arrive at our final interpretable logistic regression model:
```

```
fit_final_glm <- glm(low_bwt ~ len_preg + marital,family = binomial(), data = birt
hs)
summary(fit_final_glm)
```

```
##
## Call:
## glm(formula = low_bwt ~ len_preg + marital, family = binomial(),
##      data = births)
##
## Deviance Residuals:
##      Min        1Q    Median        3Q        Max
## -3.4403    0.1479    0.2132    0.3012    1.9834
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -23.63838     2.24101  -10.548  <2e-16 ***
## len_preg        0.70286     0.06131   11.464  <2e-16 ***
## maritalunmarried -0.66923     0.27075   -2.472   0.0134 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 692.56  on 997  degrees of freedom
## Residual deviance: 396.08  on 995  degrees of freedom
## (2 observations deleted due to missingness)
## AIC: 402.08
##
## Number of Fisher Scoring iterations: 6
```

```
# both variables have a p-value lesss than significance level 0.05,
# the difference between the null and residual deviance is good
# the degrees of freedom for hte model is very high for our sample size of 1000 ob
servations
# AIC = 402.08, hence indicating that the model is statistically significant
pR2(fit_final_glm)
```

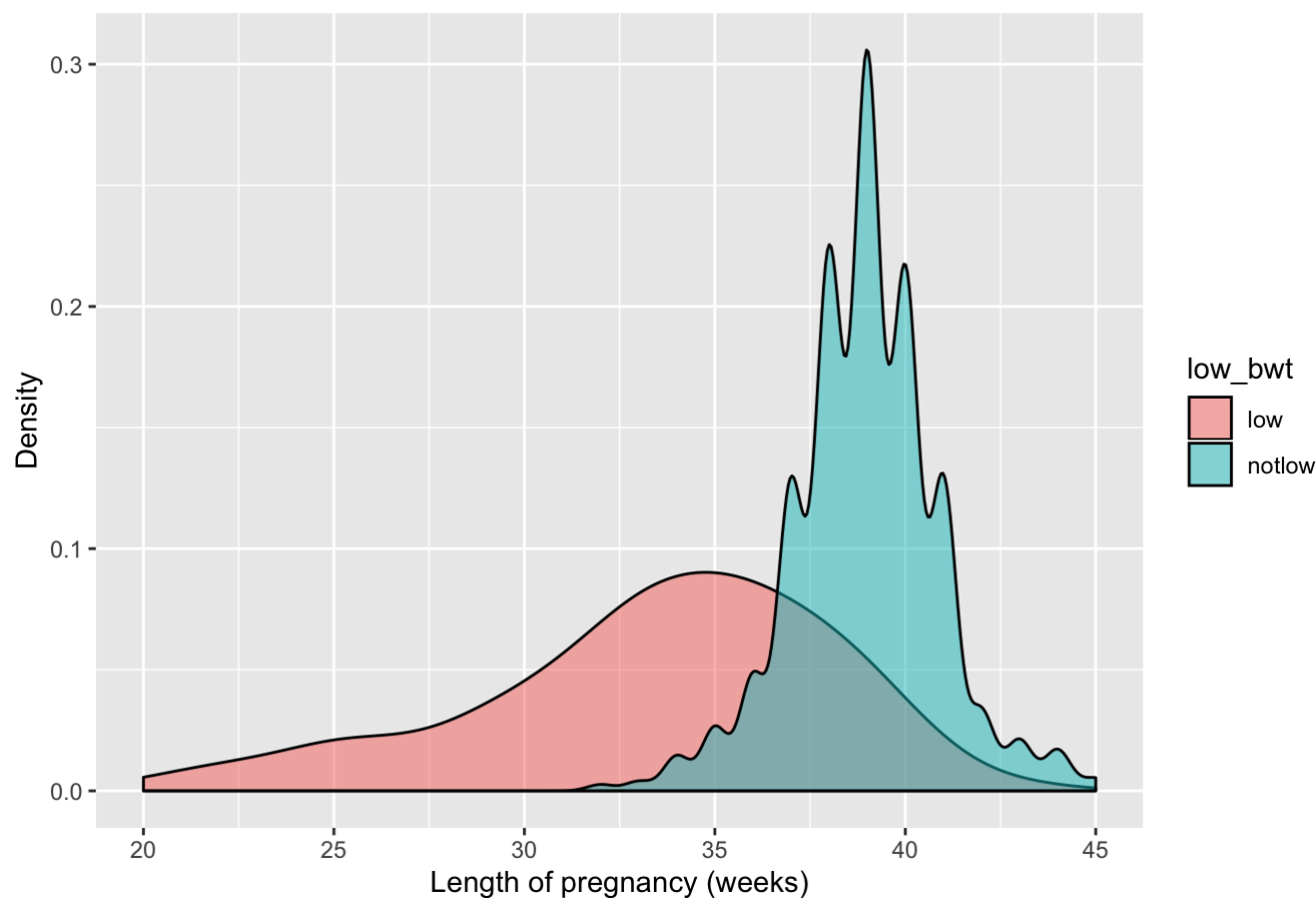
```
##          llh          llhNull          G2          McFadden          r2ML
## -198.0396885 -348.6009843  301.1225915    0.4319015    0.2604594
##          r2CU
##    0.5181017
```

```
# the McFadden value = 0.4319, also indicates a good fit
```

```
g20 = ggplot(births, aes(x=len_preg, fill=low_bwt))
g20 + geom_density(alpha=0.5) + ggtitle("Distribution of length of pregnancy and b
aby having low birthweight or not")+
  xlab ("Length of pregnancy (weeks)") + ylab("Density")
```

```
## Warning: Removed 2 rows containing non-finite values (stat_density).
```

Distribution of length of pregnancy and baby having low birthweight or not

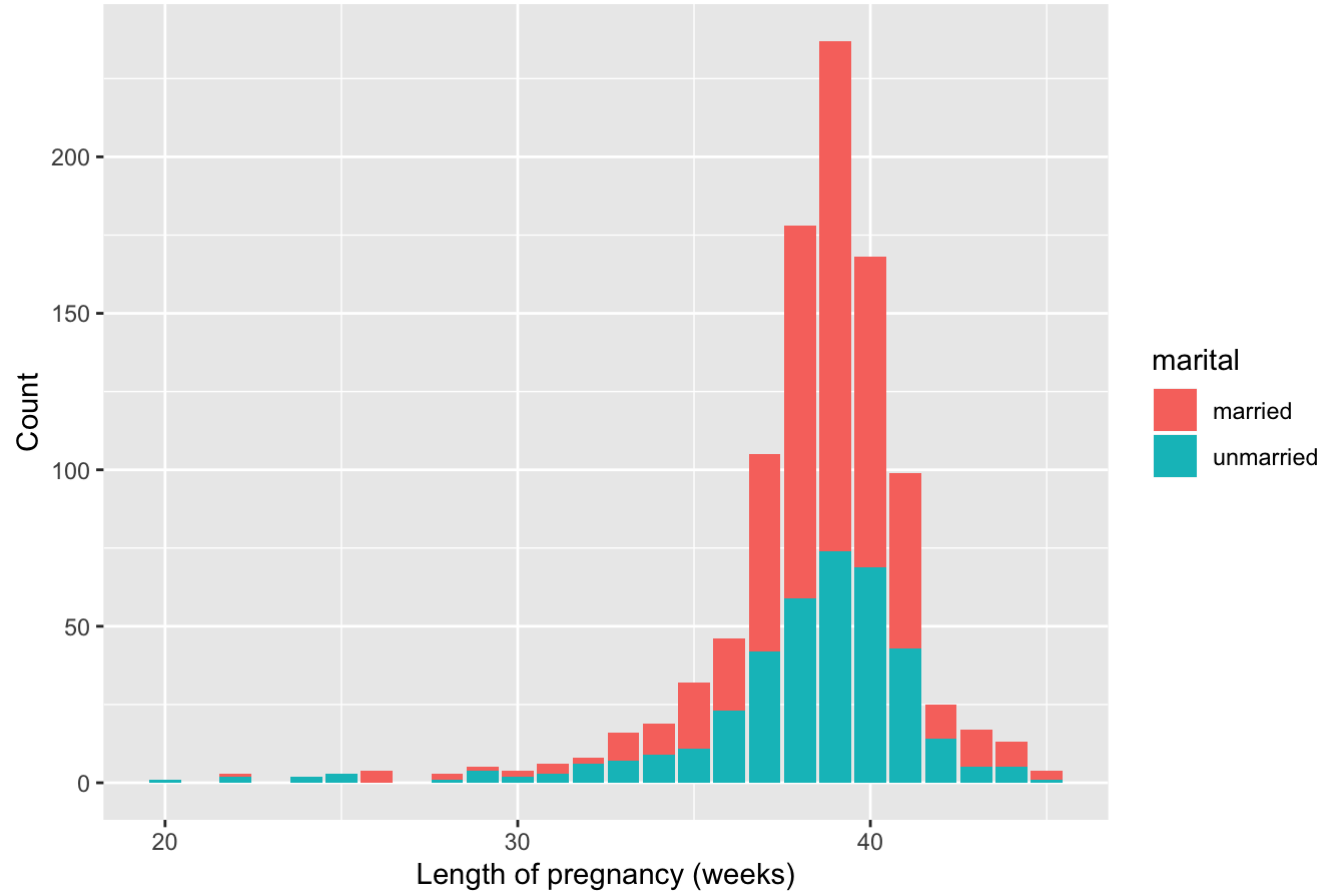


plot shows that higher the length of pregnancy, more is the chance that the baby will not have a low birthweight

```
g21 = ggplot(births, aes(x=len_preg, fill=marital))
g21 + geom_bar() + ggtitle("Distribution of length of pregnancy and mother's marital status") +
  xlab("Length of pregnancy (weeks)") + ylab("Count")
```

```
## Warning: Removed 2 rows containing non-finite values (stat_count).
```


Distribution of length of pregnancy and mother's marital status



plot reveals that married mother's tend to have longer length of pregnancy as compared to unmarried mothers.

Brief Interpretation

the probability of the baby having low birthweight or not is highly dependent on the variables:

length of pregnancy and mother's marital status

fitted equation of the logistic model:

*# low_bwt = 0.70286 * len_preg - 0.66923 * marital - 23.63838*

for len_preg:

For every 1 week increase in length of pregnancy, the log odds of having baby with low birthweight

increases by 0.091152 on average, when all other variables in the model are held constant

In the fit_final_glm: $\exp(\text{coef}(\text{len_preg})) = 2.019516$

For every 1 week increase in length of pregnancy, the odds of having baby with low birthweight is

multiplied by 0.091152 on average, when all other variables in the model are held constant

for marital:

For married as compared with unmarried mothers, the log odds of having baby with low birthweight

decreases by 0.66923 on average, when all other variables in the model are held constant

In the fit_final_glm: $\exp(\text{coef}(\text{marital})) = 0.512104$

For married as compared with unmarried mothers, the odds of having baby with low birthweight

is multiplied by 0.512104 on average, when all other variables in the model are held constant