# **Project**

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```
library(readr)
library(dplyr)
library(ggplot2)
library(knitr)
library(tidyverse)
library(car)
library(ResourceSelection)
library(pROC)
library(caret)
library (MASS)
library(pacman)
data <- read csv(file = 'chd.csv')</pre>
data$famhist<- factor(data$famhist,</pre>
                       levels = c("Absent", "Present"),
                       labels = c(0, 1))
data$chd <- factor(data$chd)</pre>
kable(head(data), format = 'latex')
```

$\overline{\mathrm{sbp}}$	tobacco	ldl	adiposity	famhist	typea	obesity	alcohol	age	chd
128	0.0	2.63	23.88	0	45	21.59	6.54	57	0
160	3.0	9.19	26.47	1	39	28.25	14.40	54	1
162	7.0	7.67	34.34	1	33	30.77	0.00	62	0
136	5.8	5.90	27.55	0	65	25.71	14.40	59	0
170	0.4	4.11	42.06	1	56	33.10	2.06	57	0
130	4.0	2.40	17.42	0	60	22.05	0.00	40	0

#### str(data)

```
spc_tbl_ [420 x 10] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
$ sbp
          : num [1:420] 128 160 162 136 170 130 150 122 132 132 ...
 $ tobacco : num [1:420] 0 3 7 5.8 0.4 4 0.18 4.18 2.8 7.2 ...
          : num [1:420] 2.63 9.19 7.67 5.9 4.11 2.4 4.14 9.05 4.79 3.65 ...
 $ 1d1
 $ adiposity: num [1:420] 23.9 26.5 34.3 27.6 42.1 ...
 $ famhist : Factor w/ 2 levels "0","1": 1 2 2 1 2 1 1 2 2 2 ...
           : num [1:420] 45 39 33 65 56 60 53 44 50 56 ...
 $ typea
 $ obesity : num [1:420] 21.6 28.2 30.8 25.7 33.1 ...
 $ alcohol : num [1:420] 6.54 14.4 0 14.4 2.06 ...
            : num [1:420] 57 54 62 59 57 40 44 52 48 34 ...
 $ age
           : Factor w/ 2 levels "0","1": 1 2 1 1 1 1 2 1 1 ...
 $ chd
 - attr(*, "spec")=
  .. cols(
      sbp = col_double(),
      tobacco = col_double(),
      ldl = col_double(),
```

```
.. adiposity = col_double(),
.. famhist = col_character(),
.. typea = col_double(),
.. obesity = col_double(),
.. alcohol = col_double(),
.. age = col_double(),
.. chd = col_double()
.. )
- attr(*, "problems")=<externalptr>
```

### **EDA**

```
# CHD prevalence
kable(table(data$chd), format = 'latex')
```

Var1	Freq
0	276
1	144

```
kable(prop.table(table(data$chd)), format = 'latex')
```

Var1	Freq
0	0.6571429
1	0.3428571

	$\operatorname{sbp}$	tobacco	ldl	adiposity	obesity	alcohol	age	typea
Mean	138.49286	3.734809	4.725786	25.461000	26.070024	17.17221	43.07381	52.971429
$\overline{SD}$	20.51642	4.688402	2.069873	7.763727	4.261565	24.61094	14.52351	9.665733
Min	101.00000	0.000000	0.980000	6.740000	14.700000	0.00000	15.00000	13.000000
Max	218.00000	31.200000	15.330000	42.490000	46.580000	147.19000	64.00000	78.000000

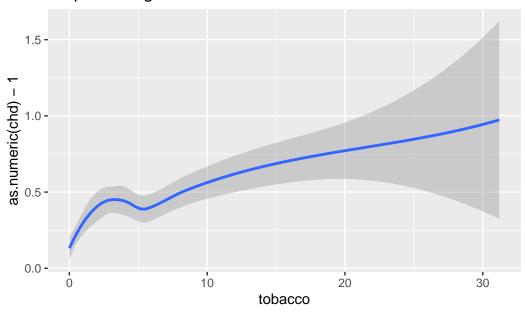
```
# Categorical variable summary
kable(table(data$famhist), format = 'latex')
```

Var1	Freq
0	243
1	177

```
# Bivariate relationships
# Empirical logit plots for continuous variables
ggplot(data, aes(x = tobacco, y = as.numeric(chd)-1)) +
   geom_smooth(method = "loess") +
   labs(title = "Empirical Logit Plot: Tobacco vs CHD")
```

`geom\_smooth()` using formula = 'y ~ x'

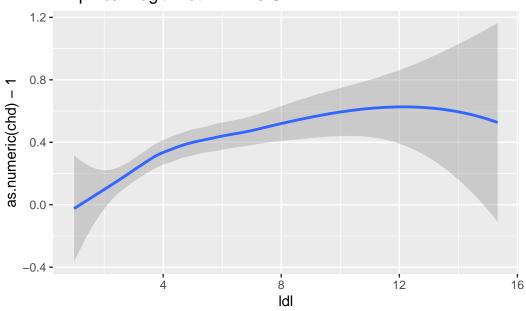
# Empirical Logit Plot: Tobacco vs CHD



```
ggplot(data, aes(x = ldl, y = as.numeric(chd)-1)) +
  geom_smooth(method = "loess") +
  labs(title = "Empirical Logit Plot: LDL vs CHD")
```

`geom\_smooth()` using formula = 'y ~ x'

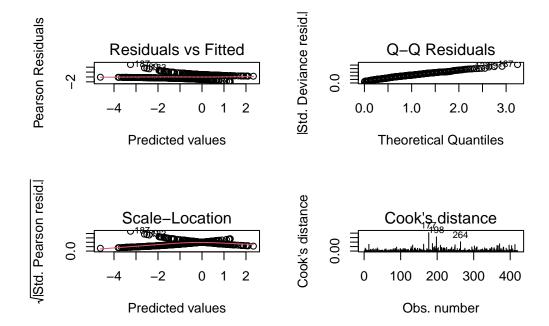
# Empirical Logit Plot: LDL vs CHD



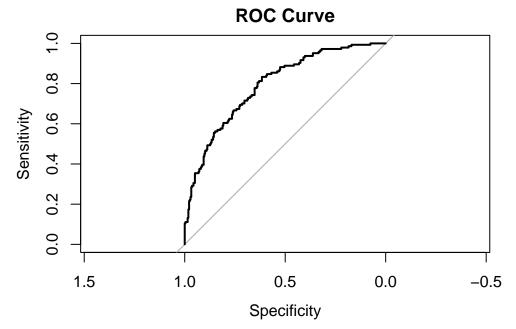
```
# Full logistic regression model
  full_model <- glm(chd ~ sbp + tobacco + ldl + adiposity + famhist +
                 typea + obesity + alcohol + age,
               family = binomial(link = "logit"), data = data)
  # Model summary
  summary(full_model)
Call:
glm(formula = chd ~ sbp + tobacco + ldl + adiposity + famhist +
   typea + obesity + alcohol + age, family = binomial(link = "logit"),
   data = data)
Coefficients:
           Estimate Std. Error z value Pr(>|z|)
(Intercept) -5.979369 1.369834 -4.365 1.27e-05 ***
                    0.005958 1.300 0.193526
sbp
           0.007746
tobacco
         ldl
adiposity 0.014893 0.030536 0.488 0.625744
famhist1 0.920423 0.239991 3.835 0.000125 ***
typea
         obesity
alcohol
         0.001774 0.004672 0.380 0.704129
         age
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 540.05 on 419 degrees of freedom
Residual deviance: 428.74 on 410 degrees of freedom
AIC: 448.74
Number of Fisher Scoring iterations: 5
  # Likelihood ratio test
  null_model <- glm(chd ~ 1, family = binomial, data = data)</pre>
  lmtest::lrtest(null_model, full_model)
Likelihood ratio test
Model 1: chd ~ 1
Model 2: chd ~ sbp + tobacco + ldl + adiposity + famhist + typea + obesity +
   alcohol + age
 #Df LogLik Df Chisq Pr(>Chisq)
  1 -270.02
2 10 -214.37 9 111.31 < 2.2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
  # McFadden's pseudo R-squared
  1 - (logLik(full_model)/logLik(null_model))
```

```
'log Lik.' 0.2061132 (df=10)
  # Stepwise model selection
  step_model <- stepAIC(full_model, direction = "both", trace = FALSE)</pre>
  summary(step model)
Call:
glm(formula = chd ~ tobacco + ldl + famhist + typea + age, family = binomial(link = "logit"),
   data = data)
Coefficients:
          Estimate Std. Error z value Pr(>|z|)
tobacco
           ldl
famhist1
           0.91050 0.23743 3.835 0.000126 ***
                    0.01272 2.197 0.028023 *
typea
           0.02794
           age
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 540.05 on 419 degrees of freedom
Residual deviance: 432.37 on 414 degrees of freedom
AIC: 444.37
Number of Fisher Scoring iterations: 5
  # Final model (based on analysis in report)
  final_model <- full_model</pre>
  # Variance Inflation Factors (check multicollinearity)
  vif(final_model)
                       ldl adiposity
     sbp
          tobacco
                                   famhist
                                                typea
                                                       obesity
                                                               alcohol
         1.156813 1.187193 3.326718 1.028312 1.069234 2.507246 1.067645
1.139224
     age
1.647164
  # Hosmer-Lemeshow test
  hoslem.test(final_model$y, fitted(final_model), g = 10)
   Hosmer and Lemeshow goodness of fit (GOF) test
data: final_model$y, fitted(final_model)
X-squared = 3.0373, df = 8, p-value = 0.932
  # Residual analysis
  par(mfrow = c(2,2))
```

plot(final\_model, which = 1:4)



```
par(mfrow = c(1,1))
  ###############################
  # Predictive Accuracy Assessment #
  ###############################
  # Predicted probabilities
  data$pred_prob <- predict(final_model, type = "response")</pre>
  # Confusion matrix (0.5 cutoff)
  # Convert both vectors to factors with EXPLICIT LEVELS
  predicted <- factor(ifelse(data$pred_prob > 0.5, "Yes", "No"),
                      levels = c("No", "Yes"))
  actual <- factor(data$chd, levels = c("No", "Yes"))</pre>
  # Create confusion matrix with aligned factors
  conf_mat <- confusionMatrix(predicted, actual)</pre>
  # ROC curve analysis
  roc_obj <- roc(data$chd, data$pred_prob)</pre>
Setting levels: control = 0, case = 1
Setting direction: controls < cases
  plot(roc_obj, main = "ROC Curve")
```



```
auc(roc_obj)
Area under the curve: 0.7932
  # Optimal cutoff using Youden's index
  coords(roc_obj, "best", best.method = "youden")
  threshold specificity sensitivity
1 0.2888442
               0.615942
                           0.8333333
  ##########################
  # Results Interpretation #
  ########################
  # Odds ratios and CI
  odds_ratios <- exp(coef(final_model))</pre>
  ci <- exp(confint(final_model))</pre>
Waiting for profiling to be done...
  results_table <- data.frame(</pre>
    Predictor = names(odds_ratios),
    OR = round(odds_ratios, 3),
    CI_Lower = round(ci[,1], 3),
    CI_Upper = round(ci[,2], 3)
  )
  print(results_table)
              Predictor
                            OR CI_Lower CI_Upper
```

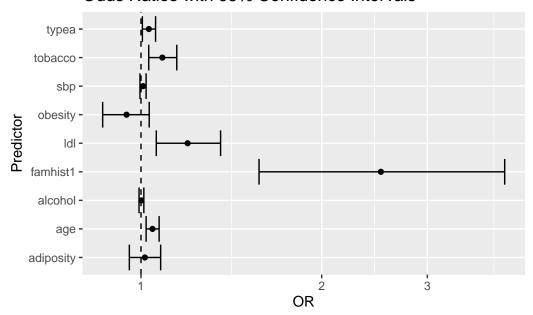
0.000

0.035

(Intercept) (Intercept) 0.003

```
0.996
                                        1.020
sbp
                   sbp 1.008
              tobacco 1.085
                               1.030
                                        1.147
tobacco
ldl
                   ldl 1.196
                               1.061
                                       1.357
            adiposity 1.015
                               0.956
                                       1.078
adiposity
famhist1
             famhist1 2.510
                               1.573
                                      4.036
typea
                 typea 1.031
                               1.005
                                        1.058
              obesity 0.946 0.863
                                       1.032
obesity
alcohol
               alcohol 1.002
                               0.993
                                       1.011
                                        1.072
age
                   age 1.045
                               1.020
  # Clinical interpretation
  cat("Key Interpretation:\n")
Key Interpretation:
  cat("- Each additional cigarette/day increases CHD odds by", round(100*(results_table["tobacco"
- Each additional cigarette/day increases CHD odds by 8.5 %
  cat("- Each unit increase in LDL increases CHD odds by", round(100*(results_table["ldl","OR"]-1
- Each unit increase in LDL increases CHD odds by 19.6 %
  cat("- Family history increases CHD odds by", round(results_table["famhistYes","OR"],1), "times
- Family history increases CHD odds by NA times
  # Visualization Code #
  # Coefficient plot
  coef_plot <- data.frame(Predictor = names(coef(final_model))[-1],</pre>
                        OR = exp(coef(final_model))[-1],
                        CI_Lower = exp(confint(final_model))[-1,1],
                        CI_Upper = exp(confint(final_model))[-1,2])
Waiting for profiling to be done...
Waiting for profiling to be done...
  ggplot(coef_plot, aes(x = OR, y = Predictor)) +
    geom_point() +
    geom_errorbarh(aes(xmin = CI_Lower, xmax = CI_Upper)) +
    geom_vline(xintercept = 1, linetype = "dashed") +
    scale_x_log10() +
    labs(title = "Odds Ratios with 95% Confidence Intervals")
```

### Odds Ratios with 95% Confidence Intervals



# Probability Distribution by CHD Status

