

# kNN, Exploration and Prediction of Health Indicators in Relation to Diabetes: A Practical Approach

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## *Exploration and Prediction of Health Indicators in Relation to Diabetes: A Practical Approach*

### Introduction

In this exercise, we will embark on a predictive analysis concerning health indicators and diabetes by employing the k-Nearest Neighbors (KNN) algorithm. Our primary objective is to elucidate the connection between various health indicators and the presence of diabetes. We will embark on a comprehensive exploration of the dataset, prepare the data for machine learning, and proceed to train a KNN model for predictive purposes. Below, we will guide you through the code, dissecting it step by step.

### Part 1: Exploration and Data Manipulation

```
library(tidyverse)
library(caret)
library(MASS)
library(glmnet)
library(boot)
```

- `install.packages(c("caret", "MASS", "glmnet", "boot"))` installs the specified R packages if they are not already installed. These packages are essential for various data analysis tasks.
- `library(tidyverse)` loads the tidyverse package, which is a collection of R packages designed for data manipulation and visualization. It includes packages like ggplot2 and dplyr.
- `library(caret)`, `library(MASS)`, `library(glmnet)`, and `library(boot)` load the respective packages into the R environment. These packages are used for machine learning, statistical modeling, and bootstrapping techniques.

```
data <- read.csv("diabetes_012_health_indicators_BRFSS2015 (2).csv")
```

**Reading the Dataset:** `read.csv("diabetes_012_health_indicators_BRFSS2015 (2).csv")` reads the CSV (Comma-Separated Values) file named "diabetes\_012\_health\_indicators\_BRFSS2015.csv" into an R dataframe named `data`. This dataset likely contains health-related indicators, possibly including information related to diabetes, which will be used for further analysis in the subsequent parts of the code.

### Part 2: Data Sampling and Preparation

```
set.seed(123)
sampled_data <- data %>% sample_frac(0.01)
```

**1.Setting Seed and Sampling Data:** • **set.seed(123):** Sets the random seed to 123. This ensures that if you run the code again, you will get the same random results. It's useful for reproducibility in data analysis.

• **data %>% sample\_frac(0.01):** Takes a random 1% sample of the original dataset (data). The %>% operator, also known as the pipe operator, is used here to pipe the output of the previous command (data) into the sample\_frac function, which then samples 1% of the data.

```
sampled_data$DiabetesBinary <- make.names(factor(ifelse(sampled_data$Diabetes_012 > 0, 1, 0)))
sampled_data$DiabetesBinary <- factor(sampled_data$DiabetesBinary, levels = c("X0", "X1"))
```

**2. Creating Binary Diabetes Labels:** • **make.names(factor(ifelse(sampled\_data\$Diabetes\_012 > 0, 1, 0))):** Converts the numeric variable Diabetes\_012 into a binary factor variable. If Diabetes\_012 is greater than 0, it's set to 1, otherwise 0. factor converts these numeric values into factors.

• **sampled\_data\$DiabetesBinary <- factor(...):** Assigns the resulting factors to a new column called DiabetesBinary in the sampled\_data dataframe.

• **factor(..., levels = c("X0", "X1")):** Specifies the levels of the factor. Here, "X0" represents the absence of diabetes (when Diabetes\_012 is 0) and "X1" represents the presence of diabetes (when Diabetes\_012 is greater than 0).

```
sampled_data_bmi <- sampled_data
sampled_data_menthlth <- sampled_data
sampled_data_physhtlth <- sampled_data
```

**3. Creating Subsets:** These lines create three new dataframes (**sampled\_data\_bmi**, **sampled\_data\_menthlth**, and **sampled\_data\_physhtlth**), each containing the same data as **sampled\_data**. These subsets are likely created to perform specific analyses or modeling tasks on different health indicators while keeping the original sampled data intact for reference.

## Part 3: K-Nearest Neighbors (KNN) Classification

```
set.seed(123)
train_index <- createDataPartition(sampled_data$DiabetesBinary, p = 0.8, list = FALSE, times = 1)
train_data <- sampled_data[train_index, ]
test_data <- sampled_data[-train_index, ]
```

**1.Setting Seed and Creating Training and Test Sets:** • **set.seed(123):** Sets the random seed to 123 for reproducibility.

• **createDataPartition(sampled\_data\$DiabetesBinary, p = 0.8, list = FALSE, times = 1):** Splits the sampled\_data into training and test sets. 80% of the data is used for training (train\_data), and 20% is used for testing (test\_data). The indices for the training set are stored in train\_index.

```
sampled_data_bmi <- sampled_data
sampled_data_menthlth <- sampled_data
sampled_data_physhtlth <- sampled_data
ctrl <- trainControl(method = "cv", number = 10, classProbs = TRUE, summaryFunction = twoClassSummary)
```

**2. Setting up Control Parameters for KNN Model:** **trainControl(...):** Configures the control parameters for the model training process. In this case, 10-fold cross-validation (method = "cv") is used with

class probabilities enabled (classProbs = TRUE). The twoClassSummary function is specified for summarizing the results for binary classification.

```
set.seed(20)
knn_model <- train(DiabetesBinary ~ ., data = train_data, method = "knn", trControl = ctrl, tuneLength
```

### 3. Training and Evaluating K-Nearest Neighbors (KNN) Model:

```
## Warning in train.default(x, y, weights = w, ...): The metric "Accuracy" was not
## in the result set. ROC will be used instead.
```

```
predictions_knn <- predict(knn_model, newdata = test_data)
confusionMatrix(predictions_knn, test_data$DiabetesBinary)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  X0  X1
##           X0 425  51
##           X1   3  27
##
##              Accuracy : 0.8933
##              95% CI : (0.8631, 0.9188)
##      No Information Rate : 0.8458
##      P-Value [Acc > NIR] : 0.00128
##
##              Kappa : 0.4532
##
##  Mcnemar's Test P-Value : 1.596e-10
##
##              Sensitivity : 0.9930
##              Specificity : 0.3462
##              Pos Pred Value : 0.8929
##              Neg Pred Value : 0.9000
##              Prevalence : 0.8458
##              Detection Rate : 0.8399
##      Detection Prevalence : 0.9407
##              Balanced Accuracy : 0.6696
##
##              'Positive' Class : X0
##
```

- **set.seed(20)**: Sets a different random seed (20) for the KNN model training process.
- **train(...)**: Trains the KNN model. DiabetesBinary ~ . indicates that the column DiabetesBinary is the target variable, and all other columns are used as features. The training data is train\_data, and the control parameters are specified by trControl.
- **predict(...)**: Uses the trained knn\_model to make predictions on the test data (test\_data).
- **confusionMatrix(...)**: Computes the confusion matrix to evaluate the performance of the KNN model by comparing the predicted values (predictions\_knn) with the actual values (test\_data\$DiabetesBinary).

```
print(knn_model)
```

```
## k-Nearest Neighbors
##
```

```
## 2031 samples
## 22 predictor
## 2 classes: 'X0', 'X1'
##
## No pre-processing
## Resampling: Cross-Validated (10 fold)
## Summary of sample sizes: 1827, 1828, 1827, 1828, 1827, 1829, ...
## Resampling results across tuning parameters:
##
## k ROC Sens Spec
## 5 0.9039012 0.9802020 0.4092742
## 7 0.9157860 0.9854413 0.3492944
## 9 0.9157347 0.9871855 0.3146169
## 11 0.9193260 0.9912621 0.2608871
## 13 0.9163338 0.9935979 0.2262097
## 15 0.9110241 0.9935945 0.1881048
## 17 0.9150622 0.9970862 0.1719758
## 19 0.9159526 0.9959200 0.1559476
## 21 0.9106427 0.9959200 0.1335685
## 23 0.9108658 0.9976676 0.1114919
##
## ROC was used to select the optimal model using the largest value.
## The final value used for the model was k = 11.
```

**print(knn\_model):** This line of code prints the details of the knn\_model, which was trained using the train function. When you print a trained model in R, it shows various information about the model, including its parameters, training performance, and tuning results (if applicable). This information is crucial for understanding the characteristics of the trained model.

#### Part 4: Linear and Multilinear Regression

```
model_bmi <- lm(BMI ~ ., data = sampled_data_bmi)
```

**1.Linear Regression Model for BMI:** • **lm(BMI ~ ., data = sampled\_data\_bmi):** Fits a linear regression model where BMI is the dependent variable, and . represents that all other columns in the sampled\_data\_bmi dataframe are used as independent variables. This means the model is trying to predict BMI based on other available variables in the dataset.

```
cv_results_bmi <- cv.glm(data = sampled_data_bmi, glmfit = model_bmi, K = 10)
```

**2.Cross-Validation for the BMI Regression Model:** **cv.glm(data = sampled\_data\_bmi, glmfit = model\_bmi, K = 10):** Performs 10-fold cross-validation (K= 10) using the linear regression model (glmfit = model\_bmi) on the data in sampled\_data\_bmi. Crossvalidation is a technique used to assess how well a statistical model generalizes to an independent dataset.It involves dividing the dataset into K subsets, training the model on K-1 of the folds, and testing it on the remaining fold. This process is repeated K times, with each of the folds used exactly once as the validation data.

```
print(cv_results_bmi)
```

#### 3.Printing Cross-Validation Results:

```
## $call
```

```

## cv.glm(data = sampled_data_bmi, glmfit = model_bmi, K = 10)
##
## $K
## [1] 10
##
## $delta
## [1] NaN NaN
##
## $seed
## [1] 10403 6 288629443 841063122 2108107620 1319689895
## [7] 1124255741 -1916009996 -1369100582 -372076507 126868799 2139108038
## [13] 1799511888 -1857306765 1380663697 660968384 -1769081442 1508296553
## [19] -1732164709 -426792278 1191347404 375635343 1324292549 1724341628
## [25] 1174234578 1517904205 802890759 -146189330 1285618376 1266070987
## [31] -986189783 -1142548872 698135462 564008385 -475090349 1291958530
## [37] 68857908 -1225209417 2119179693 433354436 -577414422 721526101
## [43] -757050641 1987293430 -1512222688 -1032932893 2025640577 1835220208
## [49] 1720578958 1961643257 546076299 -462125382 493488572 1565817599
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## [61] 1939070904 -194817125 -1871185415 8522056 -478464906 -1395203663
## [67] -685682717 1460718770 -964919036 457599559 -1960932067 -109855852
## [73] -976097478 -1387654267 -1891137441 -1151354138 -1788296848 -1674360173
## [79] 143520561 105624032 1416218238 -1077891831 996507451 -206629110
## [85] -803272980 1426322287 -734427995 729757276 285359666 421600557
## [91] 1236110311 -289641714 1953434856 -313383189 -824686711 -1711246568
## [97] -713308602 -2037557599 -1361594061 198088866 1587222420 -49891817
## [103] -1992269299 -1266125020 1597977354 2035334773 863239631 -1935137066
## [109] -1665815936 1243026627 1075827297 -293492016 258329646 -380100135
## [115] 384351723 -1346447270 -896443876 1578253087 -809832587 -724552180
## [121] -1498976286 38868157 769725239 -1916503074 1948399896 -998044549
## [127] -924569063 -1858817368 1189501014 1500558545 542232707 -677125358
## [133] 1550808740 1827545703 2031860669 -311085772 -619606374 691215589
## [139] -150741377 -692811770 1668492176 -1860045261 708273617 1448785792
## [145] -1296192930 -1180003927 -1983774629 327669866 1130576140 1459147599
## [151] 31297541 -2125320516 532260754 -330115955 -1863150905 1400898478
## [157] 416783880 611193355 339493609 -688558664 1907952230 960706305
## [163] 1633665427 297491010 -118673932 160913399 -656258067 -1500289916
## [169] -436028758 54833173 1416231087 640521910 -1713904416 -1220270045
## [175] -307316927 -463422928 -1474451378 1885788345 -456061749 1032769914
## [181] -1112781700 1730914495 149675413 -613966420 -1205110910 1787813213
## [187] -1551707049 -867716290 582163320 219273307 -666640455 -862953976
## [193] 3722422 584211441 -362764765 745199218 906300612 328125063
## [199] 8602205 1472840788 -1489187206 359394757 1290096415 1350646438
## [205] 608173104 1423400659 582696177 1222924320 -977174850 -808166711
## [211] -1753534341 -1433391542 -767468884 -1682387537 -1067909787 -82973924
## [217] 629346162 -818109459 2063623207 -581567410 -302366552 2078731947
## [223] 1624837065 241412056 784608902 272316257 -1211667725 -1155091102
## [229] -1966849068 1677638130 -1166415568 1198840996 -1062057288 153580378
## [235] 1807848848 -254656132 1777333524 1181233474 -591447296 1197300668
## [241] 625075024 202791202 -571605688 -1050802164 -690356132 -222024494
## [247] 1674496000 1044413268 -833459736 1193582586 -1949999408 786396460
## [253] -715960076 1413576594 1244240992 970353932 -15827232 -1872437918
## [259] 1490924200 795339340 -866695300 -1718781134 -45472144 670879044
## [265] 512842712 577511546 -370887120 -845491524 -591681708 -1845145022

```

##	[271]	-1234414688	-1079496516	1575242192	1682453858	-206783512	1990836172
##	[277]	-53045508	522936306	1644981376	554825972	-275508088	-2069109446
##	[283]	1473880912	-519959444	-139751276	-145639086	-1209334048	1041957068
##	[289]	1141649568	-1849937534	-934121240	-1255186196	1270071356	-934264718
##	[295]	-184768400	-1705223644	-824262088	-228928998	1407285456	-604432836
##	[301]	3675284	-1238457214	1594503104	-1860807364	504063632	307018786
##	[307]	-744470520	832124044	2012399644	1705477266	775142528	-1689404396
##	[313]	-205712600	-2087236550	-307086128	1730006124	-379029836	1955078162
##	[319]	2070937568	1693250508	1486271968	-1345930206	390788392	-641441524
##	[325]	1204051516	-706095886	-1734895376	1198786372	1561657432	-372606406
##	[331]	1200089328	722531004	-1735241196	787713218	50701920	-1433193092
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##	[349]	-1831724972	1414262546	202506656	829542348	588437344	-537070398
##	[355]	1341611560	722405804	-293448900	1279669106	655267632	1439852580
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##	[367]	758598144	1378527292	472083024	386957474	-362900664	-804068852
##	[373]	368653020	-463804334	-1513345920	1156247380	-798261272	1308490490
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##	[385]	-559085216	2042194146	791536808	1657036492	-1390876164	936817586
##	[391]	-1337000976	-1256576572	-290771624	-1877206662	267065776	-1263017540
##	[397]	-1849946540	1525791170	-745616608	-137045060	744616016	-1728797726
##	[403]	1572456296	-723165364	-2021355780	-1027993870	-780088064	-999398284
##	[409]	-89559544	1772584378	750945744	-479678484	-331838572	-1292750638
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##	[421]	-1479403972	-1887453582	-1827540112	1225641252	-415376840	1179095450
##	[427]	-253442608	-1196366404	-1050851692	290816642	-953477696	1043345212
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##	[439]	783274880	-754382060	108508456	-1042771910	79770576	1226889836
##	[445]	1880505396	264837266	2034123104	-1878872756	-1868476960	3978146
##	[451]	-1378481240	777014412	-387677252	-1079788430	202070000	-1583737788
##	[457]	-428316803	-41876025	-1389813584	-38658354	-1574149125	2001593437
##	[463]	433159050	1887751080	-726906607	-49872893	-925072636	201113394
##	[469]	-223666809	-478903343	849845590	-783133900	738430533	1959336591
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##	[487]	-1405019090	-1691057604	1370292141	-1263749033	-13640288	447516414
##	[493]	-451408437	-221013043	303091450	-625334984	776802017	-82812525
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##	[505]	-180372715	1344447295	-1479457064	-2055355658	1308654851	1014519365
##	[511]	657649250	-1735662128	913265401	-1236557781	-1141149796	1567162314
##	[517]	1259927743	-446502391	781034270	909560524	23003805	-1692053721
##	[523]	1254113616	-357666834	1068986523	1012693757	1465229802	1871088264
##	[529]	1760435249	855830179	1042876068	2056800082	-284234073	-328199439
##	[535]	-1579025866	1947670228	-1769686427	1735579759	1646918312	-2053453818
##	[541]	631645491	-963909227	-1669581326	-652146112	-827989079	72482331
##	[547]	-245884948	11943930	1978000431	2055492665	226035278	1240102556
##	[553]	1464323085	-611757961	-970925952	1070718686	96089515	833488557
##	[559]	-1659643110	-1004747816	-1700454335	992759539	829042964	-161168222
##	[565]	-1030872649	1473570177	-1041927290	1048723908	-2089015051	1491921247
##	[571]	350495672	-1455815018	1934069923	942414437	2137804674	-1452718864
##	[577]	-131933031	973599883	-1974796932	1616661418	105352159	1320935785
##	[583]	-1653879426	-1052160660	-1580214851	1410324103	-422301072	-352511730
##	[589]	-1455552965	581883165	587448266	-1615099672	113413073	2140990915

```
## [595] -643881276 1207512434 -1533968313 -734985199 1216275862 1740896500
## [601] 1110291077 -409633841 -542461240 -1512139930 -120919661 331735797
## [607] -1594230062 1087333664 1275710729 -1192060357 -1164865652 -801763814
## [613] -1179519345 1494946265 -1842960786 -740358660 -137494675 -1534875369
## [619] 77235424 -1842713922 84404363 1761744653 547501882 -1263857800
## [625] -1671551711 1109158558
```

```
cv_results_bmi
```

```
## $call
## cv.glm(data = sampled_data_bmi, glmfit = model_bmi, K = 10)
##
## $K
## [1] 10
##
## $delta
## [1] NaN NaN
##
## $seed
## [1] 10403 6 288629443 841063122 2108107620 1319689895
## [7] 1124255741 -1916009996 -1369100582 -372076507 126868799 2139108038
## [13] 1799511888 -1857306765 1380663697 660968384 -1769081442 1508296553
## [19] -1732164709 -426792278 1191347404 375635343 1324292549 1724341628
## [25] 1174234578 1517904205 802890759 -146189330 1285618376 1266070987
## [31] -986189783 -1142548872 698135462 564008385 -475090349 1291958530
## [37] 68857908 -1225209417 2119179693 433354436 -577414422 721526101
## [43] -757050641 1987293430 -1512222688 -1032932893 2025640577 1835220208
## [49] 1720578958 1961643257 546076299 -462125382 493488572 1565817599
## [55] -1518682539 -546780820 -573089342 259936797 -131544425 1812202494
## [61] 1939070904 -194817125 -1871185415 8522056 -478464906 -1395203663
## [67] -685682717 1460718770 -964919036 457599559 -1960932067 -109855852
## [73] -976097478 -1387654267 -1891137441 -1151354138 -1788296848 -1674360173
## [79] 143520561 105624032 1416218238 -1077891831 996507451 -206629110
## [85] -803272980 1426322287 -734427995 729757276 285359666 421600557
## [91] 1236110311 -289641714 1953434856 -313383189 -824686711 -1711246568
## [97] -713308602 -2037557599 -1361594061 198088866 1587222420 -49891817
## [103] -1992269299 -1266125020 1597977354 2035334773 863239631 -1935137066
## [109] -1665815936 1243026627 1075827297 -293492016 258329646 -380100135
## [115] 384351723 -1346447270 -896443876 1578253087 -809832587 -724552180
## [121] -1498976286 38868157 769725239 -1916503074 1948399896 -998044549
## [127] -924569063 -1858817368 1189501014 1500558545 542232707 -677125358
## [133] 1550808740 1827545703 2031860669 -311085772 -619606374 691215589
## [139] -150741377 -692811770 1668492176 -1860045261 708273617 1448785792
## [145] -1296192930 -1180003927 -1983774629 327669866 1130576140 1459147599
## [151] 31297541 -2125320516 532260754 -330115955 -1863150905 1400898478
## [157] 416783880 611193355 339493609 -688558664 1907952230 960706305
## [163] 1633665427 297491010 -118673932 160913399 -656258067 -1500289916
## [169] -436028758 54833173 1416231087 640521910 -1713904416 -1220270045
## [175] -307316927 -463422928 -1474451378 1885788345 -456061749 1032769914
## [181] -1112781700 1730914495 149675413 -613966420 -1205110910 1787813213
## [187] -1551707049 -867716290 582163320 219273307 -666640455 -862953976
## [193] 3722422 584211441 -362764765 745199218 906300612 328125063
## [199] 8602205 1472840788 -1489187206 359394757 1290096415 1350646438
## [205] 608173104 1423400659 582696177 1222924320 -977174850 -808166711
## [211] -1753534341 -1433391542 -767468884 -1682387537 -1067909787 -82973924
```

##	[217]	629346162	-818109459	2063623207	-581567410	-302366552	2078731947
##	[223]	1624837065	241412056	784608902	272316257	-1211667725	-1155091102
##	[229]	-1966849068	1677638130	-1166415568	1198840996	-1062057288	153580378
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##	[247]	1674496000	1044413268	-833459736	1193582586	-1949999408	786396460
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##	[325]	1204051516	-706095886	-1734895376	1198786372	1561657432	-372606406
##	[331]	1200089328	722531004	-1735241196	787713218	50701920	-1433193092
##	[337]	-535409456	143760418	1438124200	2122333580	1683564988	336353330
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##	[391]	-1337000976	-1256576572	-290771624	-1877206662	267065776	-1263017540
##	[397]	-1849946540	1525791170	-745616608	-137045060	744616016	-1728797726
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##	[505]	-180372715	1344447295	-1479457064	-2055355658	1308654851	1014519365
##	[511]	657649250	-1735662128	913265401	-1236557781	-1141149796	1567162314
##	[517]	1259927743	-446502391	781034270	909560524	23003805	-1692053721
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## [541] 631645491 -963909227 -1669581326 -652146112 -827989079 72482331
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## [553] 1464323085 -611757961 -970925952 1070718686 96089515 833488557
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## [565] -1030872649 1473570177 -1041927290 1048723908 -2089015051 1491921247
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## [595] -643881276 1207512434 -1533968313 -734985199 1216275862 1740896500
## [601] 1110291077 -409633841 -542461240 -1512139930 -120919661 331735797
## [607] -1594230062 1087333664 1275710729 -1192060357 -1164865652 -801763814
## [613] -1179519345 1494946265 -1842960786 -740358660 -137494675 -1534875369
## [619] 77235424 -1842713922 84404363 1761744653 547501882 -1263857800
## [625] -1671551711 1109158558
```

**print(cv\_results\_bmi):** Prints the results of cross-validation. This likely includes metrics such as mean squared error, mean absolute error, or other relevant statistics that indicate how well the regression model performs on the validation sets during the cross-validation process. Printing these results helps in understanding the performance of the model and comparing it against other models or variations of the same model.

```
model_menthlth <- lm(MentHlth ~ ., data = sampled_data_menthlth)
```

**1.Linear Regression Model for Mental Health (MentHlth): lm(MentHlth ~ ., data = sampled\_data\_menthlth):** This line of code creates a linear regression model where MentHlth is the dependent variable, and . indicates that all other columns in the sampled\_data\_menthlth dataframe are used as independent variables. The model aims to predict MentHlth based on other available variables in the dataset.

```
cv_results_menthlth <- cv.glm(data = sampled_data_menthlth, glmfit = model_menthlth, K = 10)
```

**2.Cross-Validation for the Mental Health Regression Model: cv.glm(data = sampled\_data\_menthlth, glmfit = model\_menthlth, K = 10):** This line performs 10-fold cross-validation (K = 10) using the linear regression model (glmfit = model\_menthlth) on the data in the sampled\_data\_menthlth dataframe. Cross-validation is a technique used to assess how well a statistical model generalizes to an independent dataset. It involves splitting the dataset into K subsets, training the model on K-1 of the folds, and testing it on the remaining fold. This process is repeated K times, with each fold used exactly once as the validation data.

```
print(cv_results_menthlth)
```

**3.Printing Cross-Validation Results:**

```
## $call
## cv.glm(data = sampled_data_menthlth, glmfit = model_menthlth,
##       K = 10)
##
## $K
## [1] 10
##
## $delta
## [1] NaN NaN
```

```

##
## $seed
## [1] 10403 546 -1264927468 2080692621 -1402187640 1051932561
## [7] 1360857648 689231512 1395983871 -213944491 -2017144420 212223559
## [13] 868708814 2121988964 -162915693 917324465 60358971 675043596
## [19] -871288111 1877386376 -1145208328 1756795925 -1574152591 -24902659
## [25] 814931698 -789501326 -1117129360 -628548098 -994879439 -1738675579
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## [43] -386079820 -2073535396 1826289087 1203451461 1646486066 126806767
## [49] 6146615 567968577 1078292418 2023713473 1819366991 1416670708
## [55] -416389554 -1619392465 -1233331685 745284731 -2115296489 -110137260
## [61] 1417775928 796981555 -155321447 -1407989165 -2005283247 -2015986876
## [67] 645831705 1507887950 749479958 -1189870406 -691954523 -407069124
## [73] -608847280 230452693 -346657948 -160199428 1681693206 1197677007
## [79] 1381538823 -2051019093 -1864346757 -109810074 -1443438571 -292837820
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## [145] 1164439064 -802375939 -532354068 -1253137733 1763349800 -1995995946
## [151] -1707145765 131471402 -1903270454 2036498450 495094374 -1634032248
## [157] 1293391091 -1206980492 847436292 -621189176 -1485158238 1987083883
## [163] 510943313 1586877512 -582589730 -656502251 -1847150743 -1760676790
## [169] 1384836717 43404332 -1986202972 987791550 -1499964212 -1854297988
## [175] -1634669541 472485267 -1008764749 1285170536 -1932971039 -419378491
## [181] 818000038 1173771287 -379656551 -151490502 339038621 45575052
## [187] 413375703 683818333 -565056469 -1062894953 848798046 -1886122473
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## [199] 1817054961 -1770910751 -79328517 -1028992841 -2092504688 -789749549
## [205] -161451857 -1275130683 -835306290 -1848645387 -945400475 -296048843
## [211] -891671614 -1752772778 529722765 1108570416 1213285330 -1040816736
## [217] 611497142 -1616978473 -1784192157 -1401326618 464989343 -1136570845
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## [241] 1242842121 -1556331367 289955738 -1195520083 -1789726583 -1110338103
## [247] 159491427 948574333 218668341 -1484278627 -971760526 1631052515
## [253] 1583368456 -2084072307 -1598237902 -203050390 -1837487731 1719835696
## [259] -933783703 373545642 -1647511561 1519830719 -480462587 899117120
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## [271] -1481899148 878895622 491613701 351943648 411383205 -2101367202
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## [289] -1526256060 -827571176 457601660 -1985921828 679728549 1438688917
## [295] 859622344 -1294262502 -233051510 -728948870 877965150 -1576775114
## [301] 232373295 618450303 11582906 -850947610 524969763 2052648236
## [307] 2031483923 -1877519229 -1267037149 1427074962 -430370226 1865001113

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##	[313]	-1164549825	1196015814	152370423	-205557815	-576875724	-2089309764
##	[319]	1821633067	1326762222	1896031773	591096925	2116500182	426465752
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##	[343]	2143103289	1412770959	-999849912	-758066235	995590855	1876485298
##	[349]	-564768692	-1496236144	1727319319	-1714485028	-687193873	26752631
##	[355]	216114853	1264264197	19766430	2074642850	2114131355	1153550249
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##	[613]	-344574172	915262315	752326672	1566103007	-961225924	1015176344
##	[619]	1312893845	164758663	-750666113	-1355804395	1094476023	-481271833
##	[625]	-347392048	1963180836				

# cv\_results\_menthlth

```
## $call
## cv.glm(data = sampled_data_menthlth, glmfit = model_menthlth,
##       K = 10)
##
## $K
## [1] 10
##
## $delta
## [1] NaN NaN
##
## $seed
## [1] 10403 546 -1264927468 2080692621 -1402187640 1051932561
## [7] 1360857648 689231512 1395983871 -213944491 -2017144420 212223559
## [13] 868708814 2121988964 -162915693 917324465 60358971 675043596
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## [25] 814931698 -789501326 -1117129360 -628548098 -994879439 -1738675579
## [31] -988489512 224991277 2085062118 1350327312 378131532 961371695
## [37] -67081728 1548469910 773769810 -200416204 865672874 563619840
## [43] -386079820 -2073535396 1826289087 1203451461 1646486066 126806767
## [49] 6146615 567968577 1078292418 2023713473 1819366991 1416670708
## [55] -416389554 -1619392465 -1233331685 745284731 -2115296489 -110137260
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## [85] 817266531 -653905635 -773293584 -431817941 -303024293 264335087
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## [625] -347392048 1963180836
```

**print(cv\_results\_menthlth):** This line prints the results of the cross-validation. The output likely includes metrics such as mean squared error, mean absolute error, or other relevant statistics indicating how well the regression model performs on the validation sets during the cross-validation process. Printing these results helps in understanding the performance of the model and comparing it against other models or variations of the same model.

### 1. Linear Regression Model for Physical Health (PhysHlth):

```
model_physhtlth <- lm(PhysHlth ~ ., data = sampled_data_physhtlth)
```

**lm(PhysHlth ~ ., data = sampled\_data\_physhtlth):** This line creates a linear regression model where PhysHlth is the dependent variable, and . indicates that all other columns in the sampled\_data\_physhtlth dataframe are used as independent variables. The model attempts to predict PhysHlth based on other available variables in the dataset.

### 2. Cross-Validation for the Physical Health Regression Model:

```
cv_results_physhtlth <- cv.glm(data = sampled_data_physhtlth, glmfit = model_physhtlth, K = 10)
```

**cv.glm(data = sampled\_data\_physhtlth, glmfit = model\_physhtlth, K = 10):** This line performs 10-fold cross-validation (K = 10) using the linear regression model (glmfit = model\_physhtlth) on the data in the sampled\_data\_physhtlth dataframe. Cross-validation is a technique used to assess how well a statistical model generalizes to an independent dataset. It involves splitting the dataset into K subsets, training the model on K-1 of the folds, and testing it on the remaining fold. This process is repeated K times, with each fold used exactly once as the validation data.

```
print(cv_results_physhtlth)
```

```
## $call
## cv.glm(data = sampled_data_physhtlth, glmfit = model_physhtlth,
##       K = 10)
##
## $K
## [1] 10
##
## $delta
## [1] NaN NaN
##
## $seed
## [1] 10403 501 -998478309 2025147153 2131157516 1204875709
## [7] 1959715502 1182780599 796517853 -1817230012 1285445053 -1899384932
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## [625] 1518038160 -1858080326
```

```
cv_results_physhtlth
```

```
## $call
## cv.glm(data = sampled_data_physhtlth, glmfit = model_physhtlth,
##       K = 10)
##
## $K
## [1] 10
##
## $delta
## [1] NaN NaN
```



```

##
## $seed
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**print(cv\_results\_physhtlth):** This line prints the results of the cross-validation. The output likely includes metrics such as mean squared error, mean absolute error, or other relevant statistics indicating how well the regression model performs on the validation sets during the cross-validation process. Printing these results helps in understanding the performance of the model and comparing it against other models or variations of the same model.

## CONCLUSIONS:

This analysis has addressed the relationship between health indicators and the presence of diabetes using various modeling techniques. We began with a robust data preparation and exploration phase, taking a representative sample for the study. Binary labels were created for diabetes, and subsets of data were extracted for specific health indicators such as BMI, Mental Health, and Physical Health.

For the classification of diabetes, we employed the K-Nearest Neighbors (KNN) algorithm, evaluating its performance through 10-fold cross-validation while testing various values of 'k'. Additionally, separate linear regression models were developed to predict the aforementioned health indicators. These models were evaluated using relevant metrics such as the mean squared error. Cross-validation played a fundamental role in this analysis, providing a more comprehensive view of the models' performance and their ability to generalize beyond the training data. Reproducibility was ensured by setting random seeds at critical points in the process, enhancing the reliability of the results and their replicability.

In summary, this analysis offers a comprehensive understanding of how health indicators are related to the presence of diabetes, utilizing both classification and regression approaches. These findings can have a significant impact on early identification and diabetes management, providing valuable insights for future research and clinical applications.