

Classification

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Pada Praktek kali ini kita akan membuat model klasifikasi dengan algoritma Decision Tree, Naive Bayes, dan K-NN menggunakan dataset insurance. Dataset ini merupakan dataset yang didapatkan dari kaggle, namun telah melalui tahapan pre-processing. sehingga data yang digunakan sudah dalam kondisi baik/siap digunakan.

Decision Tree

Import Library

```
# Import Library
library(rpart)
library(rattle)
```

```
## Warning: package 'rattle' was built under R version 3.5.3
## Rattle: A free graphical interface for data science with R.
## Version 5.2.0 Copyright (c) 2006-2018 Togaware Pty Ltd.
## Type 'rattle()' to shake, rattle, and roll your data.
```

```
library(rpart.plot)
```

```
## Warning: package 'rpart.plot' was built under R version 3.5.3
```

```
library(RColorBrewer)
```

Import Dataset

```
#Import Data
insurance <- read.csv("insurance.csv")
```

Data Exploration

```
#Melihat Kondisi Data
dim(insurance)
```

```
## [1] 1338    8
```

```
head(insurance,10)
```

	Age	Sex	Bmi	Children	Smoker	Region	Charges	Claim
## 1	19	Female	27.900	0	Smoker	Southwest	16884.924	Yes
## 2	18	Male	33.770	1	Non Smoker	Southeast	1725.552	Yes
## 3	28	Male	33.000	3	Non Smoker	Southeast	4449.462	No
## 4	33	Male	22.705	0	Non Smoker	Northwest	21984.471	No
## 5	32	Male	28.880	0	Non Smoker	Northwest	3866.855	Yes
## 6	31	Female	25.740	0	Non Smoker	Southeast	3756.622	No
## 7	46	Female	33.440	1	Non Smoker	Southeast	8240.590	Yes
## 8	37	Female	27.740	3	Non Smoker	Northwest	7281.506	No
## 9	37	Male	29.830	2	Non Smoker	Northeast	6406.411	No
## 10	60	Female	25.840	0	Non Smoker	Northwest	28923.137	No

```
#Melihat Data Kosong
sum(is.na(insurance))
```

```
## [1] 0
```

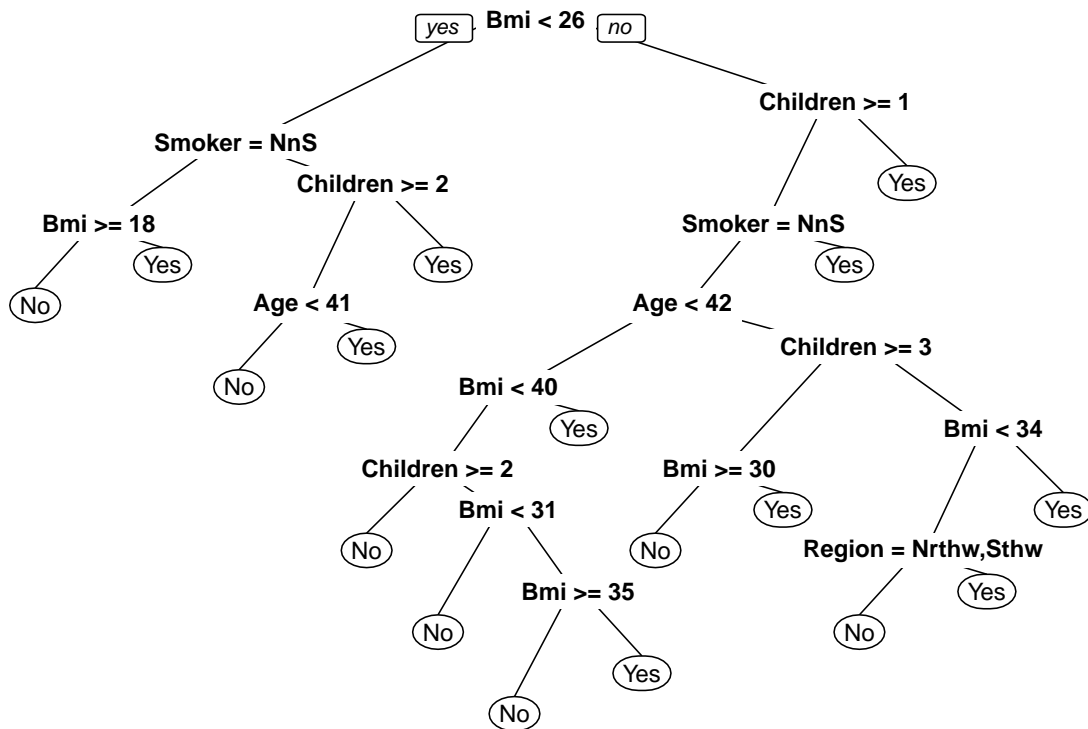
Data Preprocessing

```
#Membangi Data Ke Training dan Testing (70:30)
index_train <- sample(1:nrow(insurance), 0.7 * nrow(insurance))
train <- insurance[index_train, ]
test <- insurance[-index_train, ]
```

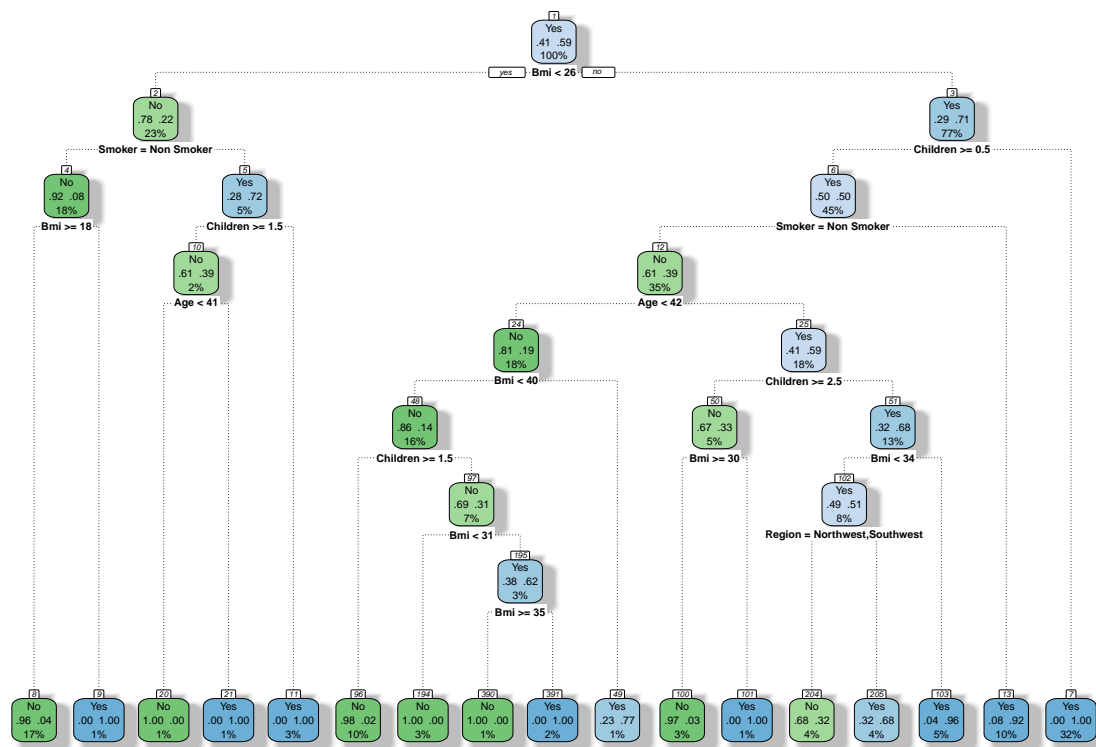
Model Building

```
#Membuat Model Decison Tree Untuk Mengklasifikasi Apakah Seseorang akan klaim Asuransi atau tidak.
tree <- rpart(Claim ~., train, method = "class")
```

```
#Memvisualisasikan Model Decison Tree
prp(tree)
```



```
#Memvisualisasikan Decison Tree dengan lebih informatif
fancyRpartPlot(tree)
```



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```
#Menggunakan Untuk Melakukan Prediksi Pada Data Testing
prediction <- predict(tree, test, type = "class")
```

Validation

```
#Validasi Menggunakan Confussion Matrix
conf <- table(test$Claim, prediction)
conf
```

```
##      prediction
##      No Yes
## No  160 15
## Yes  12 215
```

```
TP <- conf[1, 1]
FN <- conf[1, 2]
FP <- conf[2, 1]
TN <- conf[2, 2]
```

```
#Menghitung Nilai Akurasi
acc <- (TP + TN)/(TP + FN + FP + TN)
accdt <- acc
acc
```

```
## [1] 0.9328358
```

```
#Menghitung Nilai Precision
prec <- TP / (TP + FP)
```

```
prec
```

```
## [1] 0.9302326
```

```
#Menghitung Nilai Recall
```

```
rec <- TP / (TP + FN)
```

```
rec
```

```
## [1] 0.9142857
```

Naive Bayes

Import Library

```
#Import Library
```

```
library(naivebayes)
```

```
## Warning: package 'naivebayes' was built under R version 3.5.3
```

Model Building

```
#Membuat model prediksi Naive Bayes
```

```
nb <- naive_bayes(Claim ~ ., data = train)
```

```
#Melihat model yang telah dibuat
```

```
nb
```

```
## ===== Naive Bayes =====
```

```
## Call:
```

```
## naive_bayes.formula(formula = Claim ~ ., data = train)
```

```
##
```

```
## A priori probabilities:
```

```
##
```

```
##           No           Yes
```

```
## 0.4059829 0.5940171
```

```
##
```

```
## Tables:
```

```
##
```

```
## Age           No           Yes
```

```
## mean 37.53947 40.88669
```

```
## sd   12.90481 14.79078
```

```
##
```

```
##
```

```
## Sex           No           Yes
```

```
## Female 0.5394737 0.4910072
```

```
## Male   0.4605263 0.5089928
```

```
##
```

```
##
```

```
## Bmi           No           Yes
```

```
## mean 27.799618 32.775207
```

```
## sd    5.597293  5.886824
```

```
##
```

```
##
```

```
## Children      No           Yes
```

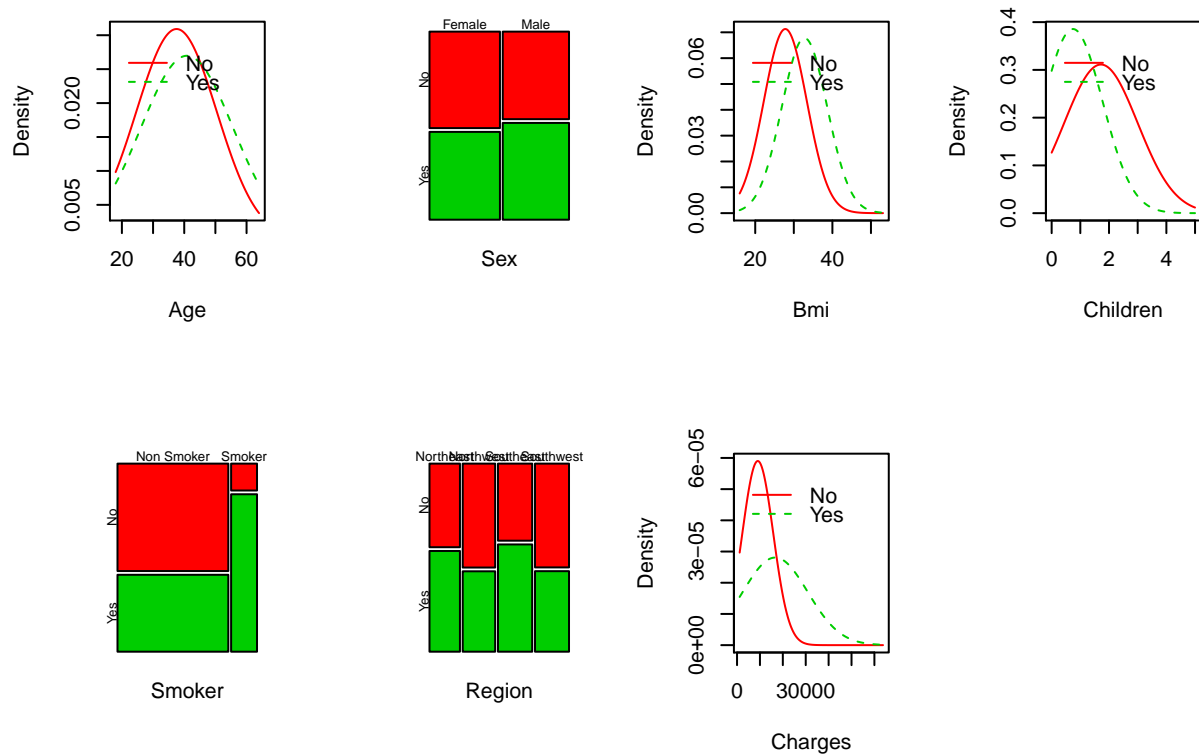
```
## mean 1.7184211 0.7428058
```

```
## sd    1.2819747 1.0347313
```

```
##
##
## Smoker           No           Yes
## Non Smoker 0.94473684 0.67625899
## Smoker      0.05526316 0.32374101
##
## # ... and 2 more tables
```

```
#Visualisasi Model
```

```
par(mfrow=c(2,4))
plot(nb)
```



```
#Melakukan prediksi dengan data testing
pred_nb <- predict(nb, as.data.frame(test))
```

Validation

```
#Membuat Confussion Matrix Naive Bayes
confnb <- table(test$Claim, pred_nb)
confnb
```

```
##      pred_nb
##      No Yes
## No  151  24
## Yes  58 169
```

```
TPn <- confnb[1, 1]
FNn <- confnb[1, 2]
```

```

FPn <- confnb[2, 1]
TNn <- confnb[2, 2]

#Menghitung Nilai Akurasi
accnb <- (TPn + TNn)/(TPn + FNn + FPn + TNn)
accnb

## [1] 0.7960199

#Menghitung Nilai Precision
precnb <- TPn / (TPn + FPn)
precnb

## [1] 0.722488

#Menghitung Nilai Recall
recnb <- TPn / (TPn + FNn)
recnb

## [1] 0.8628571

```

K-NN

Import Library

```

#import library yang dibutuhkan
library(class)
library(tidyverse)

## -- Attaching packages -----
## v ggplot2 3.1.1      v purrr  0.3.2
## v tibble  2.1.1      v dplyr  0.8.0.1
## v tidyr   0.8.3      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0

## Warning: package 'ggplot2' was built under R version 3.5.3
## Warning: package 'tibble' was built under R version 3.5.3
## Warning: package 'tidyr' was built under R version 3.5.3
## Warning: package 'purrr' was built under R version 3.5.3
## Warning: package 'dplyr' was built under R version 3.5.3

## -- Conflicts -----
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

```

Data Pre-Processing

```

#Mengubah Data Ke Tipe Numerik
insurance1 <- insurance %>% mutate_if(is.factor, as.numeric)

#Membuat fungsi Normalisasi
normalize<-function(x){
  temp<-(x-min(x))/(max(x)-min(x))
}

```

```

    return(temp)
}

#Melakukan Normalisasi
kinsurance_n<-as.data.frame(lapply(insurance1[,c(1:7)],normalize))

#Membagi ke Data Train dan Data Testing
index_train <- sample(1:nrow(kinsurance_n), 0.7 * nrow(kinsurance_n))
kinsurance_train <- kinsurance_n[index_train, ]
kinsurance_test <- kinsurance_n[-index_train, ]

#Mengambil Label
kinsurance_train_target<-insurance1[index_train,8]
kinsurance_test_target<-insurance1[-index_train, 8]

```

Model Building

```

#Membuat KNN-Model dengan Nilai K=2
knnmodel <-knn(train=kinsurance_train,test=kinsurance_test,cl=kinsurance_train_target,k=2)

```

Validation

```

#Validasi Menggunakan Confussion Matrix
confknn <- table(kinsurance_test_target, knnmodel)
confknn

```

```

##                knnmodel
## kinsurance_test_target    1    2
##                1 143   32
##                2   23  204

```

```

TPk <- confknn[1, 1]
FNk <- confknn[1, 2]
FPk <- confknn[2, 1]
TNk <- confknn[2, 2]

```

```

#Melihat Nilai Akurasi K-NN
acck <- (TPk + TNk)/(TPk + FNk + FPk + TNk)
acck

```

```
## [1] 0.8631841
```

```

#Melihat Nilai Precision K-NN
preck <- TPk / (TPk + FPk)
preck

```

```
## [1] 0.8614458
```

```

#Melihat Nilai Recall K-NN
reck <- TPk / (TPk + FNk)
reck

```

```
## [1] 0.8171429
```

Model Comparison

```
#Nilai Akurasi Decision Tree  
accdt
```

```
## [1] 0.9328358
```

```
#Nilai Akurasi Naive Bayes  
accnb
```

```
## [1] 0.7960199
```

```
#Nilai Akurasi K-NN  
acck
```

```
## [1] 0.8631841
```