Data Scientist_ML_DTS

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```
## # A tibble: 31 x 3
##
    Girth Height Volume
     <dbl> <dbl> <dbl>
##
##
  1
      8.3
             70
                 10.3
## 2
      8.6
             65 10.3
## 3 8.8
             63 10.2
## 4 10.5
             72 16.4
## 5 10.7
             81 18.8
## 6 10.8
             83 19.7
## 7 11
             66 15.6
## 8 11
             75
                 18.2
## 9 11.1
             80 22.6
## 10 11.2
            75 19.9
## # ... with 21 more rows
```

trees_df

```
names(trees_df)
## [1] "Girth" "Height" "Volume"
str(trees_df)
## spec_tbl_df [31 x 3] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ Girth : num [1:31] 8.3 8.6 8.8 10.5 10.7 10.8 11 11 11.1 11.2 ...
## $ Height: num [1:31] 70 65 63 72 81 83 66 75 80 75 ...
## $ Volume: num [1:31] 10.3 10.3 10.2 16.4 18.8 19.7 15.6 18.2 22.6 19.9 ...
## - attr(*, "spec")=
##
   .. cols(
##
   .. Girth = col_double(),
    .. Height = col_double(),
##
    .. Volume = col_double()
##
##
    ..)
## - attr(*, "problems")=<externalptr>
names(trees_df)[1] <- "Diameter"</pre>
trees_df$diameter_ft <- trees_df$Diameter*0.08333</pre>
head(trees_df)
## # A tibble: 6 x 4
## Diameter Height Volume diameter_ft
       <dbl> <dbl> <dbl>
##
                            <dbl>
               70 10.3
## 1
        8.3
                               0.692
## 2
        8.6 65 10.3
                              0.717
## 3
       8.8 63 10.2
                               0.733
        10.5
              72 16.4
## 4
                               0.875
## 5
      10.7
             81 18.8
                               0.892
## 6
      10.8
             83 19.7
                               0.900
summary(trees_df)
##
      Diameter
                      Height
                                Volume
                                             diameter_ft
## Min. : 8.30 Min. :63 Min. :10.20 Min.
                                                   :0.6916
## 1st Qu.:11.05 1st Qu.:72 1st Qu.:19.40
                                             1st Qu.:0.9208
## Median :12.90 Median :76
                              Median :24.20
                                             Median :1.0750
## Mean :13.25 Mean :76
                              Mean :30.17
                                             Mean :1.1040
## 3rd Qu.:15.25
                  3rd Qu.:80
                              3rd Qu.:37.30
                                             3rd Qu.:1.2708
## Max. :20.60 Max. :87
                              Max. :77.00
                                             Max. :1.7166
##Shapiro Test
shapiro.test(trees_df$diameter_ft)
##
## Shapiro-Wilk normality test
## data: trees_df$diameter_ft
## W = 0.94117, p-value = 0.08893
```

```
shapiro.test(trees_df$Height)

##

## Shapiro-Wilk normality test

##

## data: trees_df$Height

## W = 0.96545, p-value = 0.4034

shapiro.test(trees_df$Volume)

##

## Shapiro-Wilk normality test

##

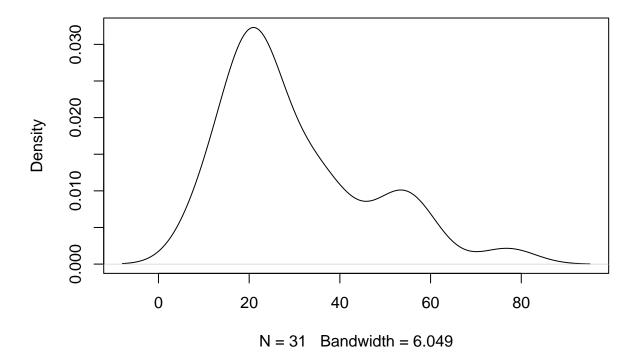
## data: trees_df$Volume

## w = 0.88757, p-value = 0.003579

#Visualisasi distribusi Volume

plot(density(trees_df$Volume))
```

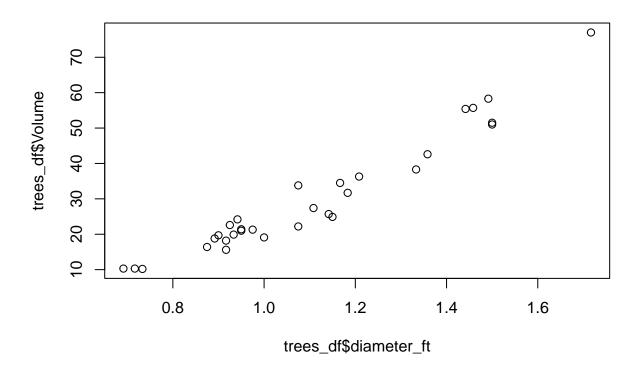
density.default(x = trees_df\$Volume)



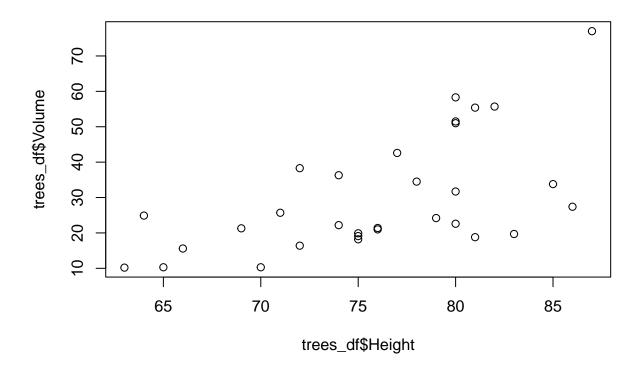
```
##Mencari hubungan
lm(formula = Volume ~ Height + diameter_ft, data = trees_df)
```

##

plot(trees_df\$diameter_ft, trees_df\$Volume)



plot(trees_df\$Height,trees_df\$Volume)



Penggunaan Machine Learning dalam analisis data terkait biaya listrik rumah tangga

```
#read dataset
electric_bill <- read.csv("https://storage.googleapis.com/dqlab-dataset/electric_bill.csv")</pre>
model <- lm(amount_paid ~ num_people+housearea, data = electric_bill)</pre>
model
##
## Call:
## lm(formula = amount_paid ~ num_people + housearea, data = electric_bill)
##
## Coefficients:
##
   (Intercept)
                  num_people
                                 housearea
       482.920
                        4.834
                                      0.118
##
Analisa Data dengan Decision Tree
iris <- read.csv("https://storage.googleapis.com/dqlab-dataset/iris.csv")</pre>
#memecah dara train dan test
trainIndex <- createDataPartition(iris$Species, p=0.8, list=FALSE)</pre>
training_set <- iris[trainIndex, ]</pre>
testing_set <- iris[-trainIndex, ]</pre>
dim(training_set)
```

```
## [1] 120 5
```

```
dim(testing_set)
## [1] 30 5
#membuat model decison tree
set.seed(123)
model_dt <- rpart(Species ~., data = training_set, method = "class")</pre>
prediction_dt <- predict(model_dt, newdata = testing_set, type = "class")</pre>
#evaluasi model dengan data test baru
testing_species = factor(testing_set$Species)
#memperlihatkan hasil evaluasi
eval_result <- confusionMatrix(prediction_dt, testing_species)</pre>
eval_result
## Confusion Matrix and Statistics
##
##
               Reference
## Prediction
              setosa versicolor virginica
##
     setosa
                    10
                                0
                                           0
##
     versicolor
                     0
                               10
                                           1
##
     virginica
                     0
                                0
                                           9
## Overall Statistics
##
##
                  Accuracy: 0.9667
##
                    95% CI: (0.8278, 0.9992)
       No Information Rate : 0.3333
##
       P-Value [Acc > NIR] : 2.963e-13
##
##
##
                     Kappa : 0.95
##
## Mcnemar's Test P-Value : NA
## Statistics by Class:
##
##
                        Class: setosa Class: versicolor Class: virginica
## Sensitivity
                               1.0000
                                                  1.0000
                                                                   0.9000
## Specificity
                               1.0000
                                                  0.9500
                                                                   1.0000
## Pos Pred Value
                               1.0000
                                                  0.9091
                                                                   1.0000
## Neg Pred Value
                               1.0000
                                                  1.0000
                                                                   0.9524
## Prevalence
                               0.3333
                                                  0.3333
                                                                   0.3333
## Detection Rate
                               0.3333
                                                  0.3333
                                                                   0.3000
## Detection Prevalence
                               0.3333
                                                  0.3667
                                                                   0.3000
## Balanced Accuracy
                               1.0000
                                                  0.9750
                                                                   0.9500
```