#### **REGRESSION TREE**

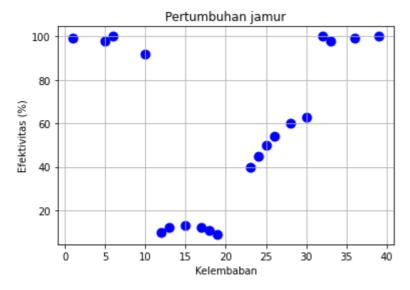
Menentukan data

```
import numpy as np
    # variabel X harus dalam bentuk array 2D
    X = np.array([[1, 5, 6, 10, 12, 13, 15, 17, 18, 19, 23, 24, 25, 26, 28, 30, 32,33, 3
    # variabel y dalam bentuk array 1D
    y = np.array([99, 98, 100, 92, 10, 12, 13, 12, 11, 9, 40, 45, 50, 54, 60, 63, 100, 9)
```

Plot data

```
import matplotlib.pyplot as plt #library untuk plot

plt.scatter(X,y, color='blue', lw=4)
  plt.xlabel("Kelembaban")
  plt.ylabel("Efektivitas (%)")
  plt.title("Pertumbuhan jamur")
  #plt.legend(('data',), loc='center left')
  plt.grid()
  plt.show()
```



## Menentukan akar dari pohon

Pada langkah ini akan diberikan contoh menentukan akar dari pohon

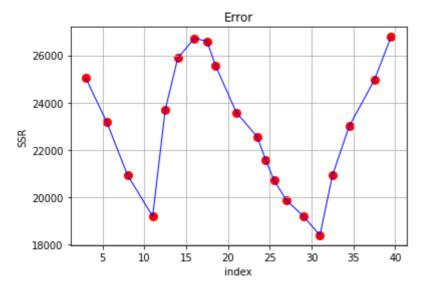
```
In [40]:
    m = len(y)
    SSR =np.array([])
    xssr = np.array([])
    for i in range(0,m):

        if (i==m-1):
        ybar = np.average(y)
        SSR = np.append(SSR, np.sum((y -ybar)**2))
        xssr = np.append(xssr,(X[i]+ (X[i]+1))/2)
        else:
        yn1 = y[0:i+1]
        ybar1 = np.average(yn1)
        yn2 = y[i+1:m]
        ybar2 = np.average(yn2)
```

```
SSR = np.append(SSR, np.sum((yn1-ybar1)**2) + np.sum((yn2 -ybar2)**2))
xssr = np.append(xssr,(X[i]+X[i+1])/2)
```

#### Ploting SSR

```
plt.scatter(xssr,SSR, color='red', lw=3)
plt.plot(xssr,SSR, color='blue', lw=1)
plt.xlabel("index")
plt.ylabel("SSR")
plt.title("Error")
plt.grid()
plt.show()
```



Mencari minimum nilai SSR

Mencari MSE dan nilai rataan akar

```
In [43]:
    m = len(y)
    value = np.average(y)
    print('Value =', value)
    MSE = np.sum((y-value)**2)/m
    print('MSE =', MSE)
```

Value = 58.25 MSE = 1339.0875

Jadi akar dan anakan pohon yang dibentuk akan berbentuk berikut ini:

# MSE = 1339.0875 Samples = 20 Value/average = 58.25 MSE = 1148.375 Sample = 16 Value/average = 48

## **Dengan SKLEARN**

```
In [44]: # import the regressor
    from sklearn.tree import DecisionTreeRegressor

# create a regressor object
    regressor = DecisionTreeRegressor(random_state = 0, min_samples_split=5)

# fit the regressor with X and Y data
    regressor.fit(X, y)

Out[44]: DecisionTreeRegressor(min_samples_split=5, random_state=0)

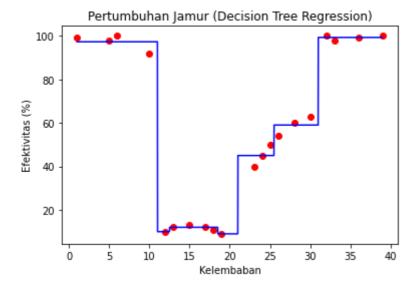
Plot Data dan Regresi Tree

In [45]: import matplotlib.pyplot as plt
```

```
import matplotlib.pyplot as plt

X_grid = np.arange(min(X), max(X), 0.01)
X_grid = X_grid.reshape((len(X_grid), 1))

plt.scatter(X, y, color = 'red')
plt.plot(X_grid, regressor.predict(X_grid), color = 'blue')
plt.title('Pertumbuhan Jamur (Decision Tree Regression)')
plt.xlabel('Kelembaban')
plt.ylabel('Efektivitas (%)')
plt.show()
```



Menyimpan dala dalam format dot untuk visualisasi

```
In [46]:
          # import export_graphviz
          from sklearn.tree import export_graphviz
          export_graphviz(regressor, out_file ='tree.dot',
                         feature_names =['Kelembaban'])
```

Plot graph Tree

```
In [47]:
              import graphviz
              graphviz.Source.from_file('tree.dot')
                                                                Kelembaban <= 31.0
squared_error = 1339.087
Out[47]:
                                                                      samples = 20
                                                                      value = 58.25
                                                                True
                                                                                     False
                                                    Kelembaban <= 11.0
                                                                                squared_error = 0.688
                                                 squared error = 1148.375
                                                                                     samples = 4
                                                       samples = 16
                                                                                     value = 99.25
                                                        value = 48.0
                                                                  Kelembaban <= 21.0
                                     squared_error = 9.688
                                                                 squared_error = 449.91
                                         samples = 4
                                                                     \overline{\text{samples}} = 12
                                         value = 97.25
                                                                     value = 31.583
                                                                                Kelembaban <= 25.5
                                                    Kelembaban <= 18.5
                                                                               squared_error = 64.333
                                                   squared_error = 1.806
                                                                                    samples = 6
                                                        samples = 6
                                                       value = 11.167
                                                                                    value = 52.0
                          Kelembaban <= 12.5
                                                     squared_error = 0.0
                                                                               squared_error = 16.667
                                                                                                            squared_error = 14.0
                          squared error = 1.04
                                                        samples = 1
                                                                                    samples = 3
                                                                                                                samples = 3
                              samples = 5
                                                         value = 9.0
                                                                                    value = 45.0
                                                                                                                value = 59.0
                              value = 11.6
              squared error = 0.0
                                       squared error = 0.5
                  samples = 1
                                           samples = 4
                  value = 10.0
                                           value = 12.0
```

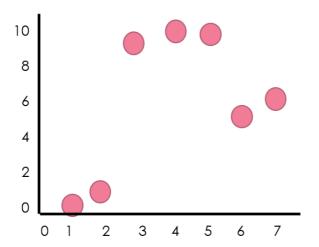
Contoh memprediksi

```
In [48]: y_pred = regressor.predict([[10]])
    print("Prediksi efektivitas: % d\n"% y_pred)
```

Prediksi efektivitas: 97

#### **LATIHAN**

Tentukan Root Node dari data berikut ini menggunakan Python



X	У
1	0
2	1
3	9
4	10
5	9.5
6	5
7	6

## **Machine Learning Approach**

Langkah 1. Menyiapkan data

Pada langkah ini menyiapkan data advertising.

```
import pandas as pd

url = 'http://bit.ly/Test-PHN'
data = pd.read_csv(url, index_col=0)

data
```

0	T 407
()IIT	1 44 1
000	172

	TV	radio	newspaper	sales
1	230.1	37.8	69.2	22.1
2	44.5	39.3	45.1	10.4
3	17.2	45.9	69.3	9.3
4	151.5	41.3	58.5	18.5
5	180.8	10.8	58.4	12.9
•••				
196	38.2	3.7	13.8	7.6
197	94.2	4.9	8.1	9.7
198	177.0	9.3	6.4	12.8
199	283.6	42.0	66.2	25.5
200	232.1	8.6	8.7	13.4

200 rows × 4 columns

Langkah 2. Membagi data menjadi 80\% training dan 20\% testing

```
import numpy as np
msk = np.random.rand(len(data)) < 0.8
train = data[msk]
test = data[~msk]
test.head()</pre>
```

```
Out[50]:
                  TV
                      radio newspaper sales
           14
                 97.5
                         7.6
                                      7.2
                                            9.7
           15 204.1
                                     46.0
                                           19.0
                        32.9
                 62.3
                        12.6
                                     18.3
                                            9.7
                                      0.3
              265.6
                        20.0
                                           17.4
           34
```

22.3

**Langkah 3**. Menyiapkan data x (TV) dan y(sales)

31.6

16.6

```
In [51]: m = len(train.TV) #number of rows data

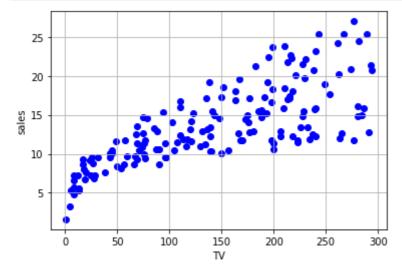
X = np.asanyarray(train[['TV']])
y = np.asanyarray(train[['sales']])
```

plot sebaran data

**41** 202.5

```
import matplotlib.pyplot as plt #library untuk plot

plt.scatter(X,y, color='blue')
plt.xlabel("TV")
plt.ylabel("sales")
plt.grid()
plt.show()
```



Langkah 4. Menentukan model yang akan digunakan.

Menggunakan regression Tree

```
In [53]: # import the regressor
```

```
from sklearn.tree import DecisionTreeRegressor

# create a regressor object
regressor = DecisionTreeRegressor(random_state = 0, min_samples_split=15)

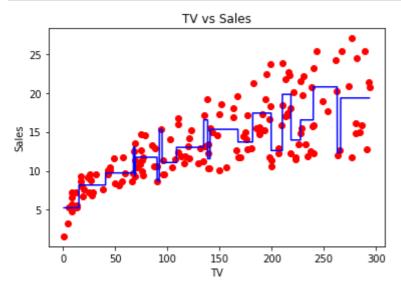
# fit the regressor with X and Y data
regressor.fit(X, y)
```

Out[53]: DecisionTreeRegressor(min\_samples\_split=15, random\_state=0)

```
In [54]:
    import matplotlib.pyplot as plt

X_grid = np.arange(min(X), max(X), 0.01)
    X_grid = X_grid.reshape((len(X_grid), 1))

plt.scatter(X, y, color = 'red')
    plt.plot(X_grid, regressor.predict(X_grid), color = 'blue')
    plt.title('TV vs Sales')
    plt.xlabel('TV')
    plt.ylabel('Sales')
    plt.show()
```



Langkah 5. Mengevaluasi model

Menentukan tabel baru yang berisi data latih/testing

```
In [55]:
    X = np.asanyarray(test[['TV']])
    y = np.asanyarray(test[['sales']])
    y = y.reshape(-1)

    print(y)

    y_pred = np.array([])
    for i in range(0,len(y)):
        y_pred = np.append(y_pred, regressor.predict([X[i]]))

    print(y_pred)
```

```
[ 9.7 19. 9.7 17.4 16.6 8.5 10.7 5.5 18. 18.9 12. 19.4 22.2 23.8 19.8 21.8 15.9 6.6 7. 11.6 8.8 24.7 19.6 10.8 20.7 13.2 10.9 19. 14.5 20.2 26.2 22.6 10.8 5.9 9.7 25.5]
[11.066666667 12.6 9.73333333 12.35 12.6 8.17692308
```

```
11.066666675.2613.0153846216.5311.7428571419.8111111120.7714285719.3461538520.7714285720.7714285713.015384628.176923088.1769230813.9511.7428571413.9513.958.1769230817.4076923120.771428578.1769230813.715.3272727320.7714285719.3461538512.68.176923088.1769230815.319.34615385]
```

Find MSE training data

```
In [56]:
MSE = np.sum((y-y_pred)**2)/len(y)
print(MSE)
```

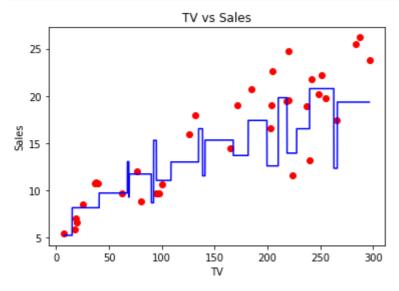
18.136070359244933

Plot test data and regression

```
import matplotlib.pyplot as plt

X_grid = np.arange(min(X), max(X), 0.01)
X_grid = X_grid.reshape((len(X_grid), 1))

plt.scatter(X, y, color = 'red')
plt.plot(X_grid, regressor.predict(X_grid), color = 'blue')
plt.title('TV vs Sales')
plt.xlabel('TV')
plt.ylabel('Sales')
plt.show()
```



## Homework

- 1. Gunakan data Advertising (X=**radio** dan y=sales) untuk memodelkan Regresi pohon dalam pendekatan Machine Learning, evaluasi hitung MSE dan plot regresinya.
- 2. Gunakan data Advertising X=**TV dan radio**, y = sales untuk memodelkan Regresi pohon dalam Machine Learning, serta hitung MSEnya.
- 3. Gunakan data Advertising X=**TV,radio dan newspaper**, y =sales untuk memodelkan Regresi pohon dalam Machine Learning, serta hitung MSEnya.

### REGRESSION TREE

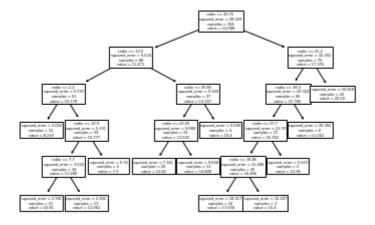
1. Gunakan data Advertising (X=radio dan y=sales) untuk memodelkan Regresi pohon

#### dalam pendekatan Machine Learning, evaluasi hitung MSE dan plot regresinya

```
In [58]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from sklearn.tree import DecisionTreeRegressor
          from sklearn import tree
In [59]:
          url = 'http://bit.ly/Test-PHN'
          data = pd.read_csv(url, index_col=0)
In [60]:
          msk = np.random.rand(len(data)) < 0.8</pre>
          train = data[msk]
          test = data[~msk]
          m = len(train.radio) #number of rows data
          X = np.asanyarray(train[['radio']])
          y = np.asanyarray(train[['sales']])
          regressor = DecisionTreeRegressor(random_state = 0, min_samples_split=33)
In [61]:
          # fit the regressor with X and Y data
          regressor.fit(X, y)
          X_{grid} = np.arange(min(X), max(X), 0.01)
          X_grid = X_grid.reshape((len(X_grid), 1))
          plt.scatter(X, y, color = 'orange')
          plt.plot(X_grid, regressor.predict(X_grid), color = 'red')
          plt.title('Plot Radio and Sales')
          plt.xlabel('Radio')
          plt.ylabel('Sales')
          plt.show()
```

## 

```
In [62]:
    tree.plot_tree(regressor,feature_names=["radio"])
    X = np.asanyarray(test[['radio']])
    y = np.asanyarray(test[['sales']])
    y = y.reshape(-1)
    y_pred = np.array([])
    for i in range(0,len(y)):
        y_pred = np.append(y_pred,regressor.predict([X[i]]) )
```

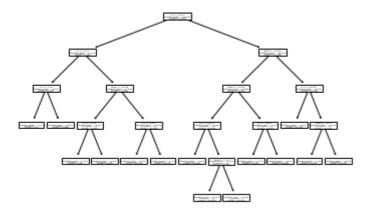


```
In [63]:
          MSE = np.sum((y-y_pred)**2)/len(y)
          print("MSE : ",MSE)
```

MSE: 19.282841879132906

# 2. Gunakan data Advertising X=TV dan radio, y = sales untuk memodelkan Regresi pohon

```
dalam Machine Learning, serta hitung MSEnya
In [64]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from sklearn.tree import DecisionTreeRegressor
          from sklearn import tree
In [65]:
          url = 'http://bit.ly/Test-PHN'
          data = pd.read_csv(url, index_col=0)
In [66]:
          msk = np.random.rand(len(data)) < 0.8</pre>
          train = data[msk]
          test = data[~msk]
          m = len(train.radio) #number of rows data
          X= np.asanyarray(train[['radio',"TV"]])
          y = np.asanyarray(train[['sales']])
          regressor = DecisionTreeRegressor(random_state = 0, min_samples_split=20)
In [67]:
          # fit the regressor with X and Y data
          regressor.fit(X, y)
          tree.plot_tree(regressor,feature_names=["radio","TV"])
          X = np.asanyarray(test[['radio',"TV"]])
          y = np.asanyarray(test[['sales']])
          y = y.reshape(-1)
          y_pred = np.array([])
          for i in range(0,len(y)):
              y_pred = np.append(y_pred,regressor.predict([X[i]]))
```

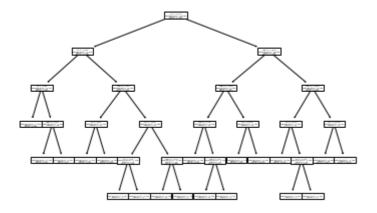


```
In [68]:
          MSE = np.sum((y-y_pred)**2)/len(y)
          print("MSE : ",MSE)
```

MSE: 1.5460102681699355

#### 3. Gunakan data Advertising X=TV,radio dan newspaper, y =sales untuk memodelkan Regresi pohon dalam Machine Learning, serta hitung MSEnya

```
In [69]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from sklearn.tree import DecisionTreeRegressor
          from sklearn import tree
In [70]:
          url = 'http://bit.ly/Test-PHN'
          data = pd.read_csv(url, index_col=0)
In [71]:
          msk = np.random.rand(len(data)) < 0.8</pre>
          train = data[msk]
          test = data[~msk]
          m = len(train.radio) #number of rows data
          X= np.asanyarray(train[['radio',"TV","newspaper"]])
          y = np.asanyarray(train[['sales']])
          regressor = DecisionTreeRegressor(random_state = 0, min_samples_split=15)
In [72]:
          # fit the regressor with X and Y data
          regressor.fit(X, y)
          tree.plot_tree(regressor,feature_names=["radio","TV","newspaper"])
          X = np.asanyarray(test[['radio',"TV","newspaper"]])
          y = np.asanyarray(test[['sales']])
          y = y.reshape(-1)
          y_pred = np.array([])
          for i in range(0,len(y)):
              y_pred = np.append(y_pred,regressor.predict([X[i]]))
```



```
In [73]:
    MSE = np.sum((y-y_pred)**2)/len(y)
    print("MSE : ",MSE)
```

MSE : 2.4608699941318957