In [1]: import numpy as np import pandas as pd import matplotlib.pyplot as plt In [2]: data = pd.read csv(r"C:\Users\ASUS\Downloads\aluminum wire rod synthetic.csv") In [3]: data.head() Out[3]: CastingTemperature(C) RollingSpeed(m/min) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) Chemical 0 517.738762 2.670609 2.124924 97.462870 1.343437 1 516.073516 4.669137 1.951931 96.631344 0.341770 2 514.097405 97.577301 0.732053 4.114840 2.440798 3 458.507464 4.946501 1.770175 98.004922 1.253575 4 502.823147 99.038671 0.430471 3.941969 0.611059 data.tail() In [4]: Out[4]: CastingTemperature(C) RollingSpeed(m/min) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComposition(Cu%) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComp 9995 450.896112 1.591570 2.469588 95.141012 1.371089 9996 454.236645 2.620234 2.536631 99.230694 0.268305 9997 496.007223 2.725959 2.837583 97.023051 1.240611 1.087624 9998 503.590094 1.449964 1.766031 96.396007 9999 472.481583 4.524516 2.319007 99.419067 1.365180 4 In [5]: data.sample(15) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) CastingTemperature(C) RollingSpeed(m/min) Chemi 2031 512.100212 4.114984 0.768487 95.415253 1.249494 1.012814 7064 97.346390 486 046415 1 011904 1 861370 7177 493.002795 3.020417 0.992113 95.852145 1.368790 502.335007 2.285010 0.829985 95.334744 1.297320 4422 9645 472.270501 3.634497 1.887266 97.555677 0.290230 0.360098 484.597591 4.987751 1996 1.433072 98.792771 8570 471.282078 3.641708 1.422238 96.787631 0.221290 9201 496.786162 2.572870 2.167432 98.067174 0.964551 3732 467.882268 3.042995 1.040646 98.449575 0.171660 8766 519.467155 1.444343 2.242932 95.517861 0.547453 8170 475.008006 4.983337 1.675417 95.330183 0.539358 5110 492.613976 4.172112 1.942683 96.487123 0.174866 9144 493.350243 4.646407 2.471678 96.558615 0.167844 489.013899 97.227616 0.441434 9215 1.549020 0.956857 475.001149 1.039078 8051 4.200922 2.753840 96.067132 data.info() In [6]:

```
#
                        Column
                                                                                   Non-Null Count Dtype
                                                                                    _ _ _ _ _ _ _ _ _ _ _ _ _ _ _ _
                 0
                         CastingTemperature(C)
                                                                                   10000 non-null
                                                                                                                 float64
                         RollingSpeed(m/min)
                 1
                                                                                   10000 non-null
                                                                                                                  float64
                 2
                         CoolingRate(C/s)
                                                                                   10000 non-null
                                                                                                                   float64
                 3
                         ChemicalComposition(Al%)
                                                                                   10000 non-null
                                                                                                                   float64
                 4
                         ChemicalComposition(Cu%)
                                                                                   10000 non-null
                                                                                                                  float64
                         ChemicalComposition(Mg%)
                                                                                   10000 non-null
                                                                                                                   float64
                 6
                         ChemicalComposition(Si%)
                                                                                   10000 non-null
                                                                                                                   float64
                 7
                         ChemicalComposition(Fe%)
                                                                                   10000 non-null
                                                                                                                   float64
                 8
                         CastBarEntryTemperature(C)
                                                                                   10000 non-null
                                                                                                                   float64
                         EmulsionTemperature(C)
                                                                                   10000 non-null
                                                                                                                   float64
                 10
                         AgingTemperature(C)
                                                                                   10000 non-null
                                                                                                                   float64
                 11
                        EmulsionPressure(Bar)
                                                                                   10000 non-null
                                                                                                                   float64
                                                                                   10000 non-null
                 12 EmulsionConcentration(%)
                                                                                                                  float64
                 13
                         RodQuenchWaterPressure(Bar)
                                                                                   10000 non-null
                                                                                                                   float64
                 14
                         FirstRollingTemperature(C)
                                                                                   10000 non-null
                                                                                                                   float64
                 15
                         QuenchingTemperature(C)
                                                                                   10000 non-null
                                                                                                                   float64
                 16
                        Elongation(%)
                                                                                   10000 non-null
                                                                                                                   float64
                  17
                         Conductivity(%IACS)
                                                                                   10000 non-null
                                                                                                                   float64
                 18 UTS(Mpa)
                                                                                   10000 non-null
                                                                                                                   float64
               dtypes: float64(19)
               memory usage: 1.4 MB
 In [7]: data.describe()
                              CastingTemperature(C) RollingSpeed(m/min) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComp
                                              10000.000000
                                                                                  10000.000000
                                                                                                                10000.000000
                                                                                                                                                               10000.000000
                                                                                                                                                                                                               10000.000000
                  count
                                                  485.100301
                                                                                          2.999822
                                                                                                                                                                     97.236129
                                                                                                                                                                                                                      0.797193
                  mean
                                                                                                                        1.741335
                      std
                                                   20.200657
                                                                                         1.160946
                                                                                                                       0.724717
                                                                                                                                                                      1.302841
                                                                                                                                                                                                                      0.403958
                                                                                                                                                                                                                      0.100020
                                                  450.005874
                                                                                          1.000230
                                                                                                                        0.500046
                                                                                                                                                                    95.000018
                     min
                    25%
                                                 467.728565
                                                                                          1.989422
                                                                                                                        1.111139
                                                                                                                                                                     96.099753
                                                                                                                                                                                                                      0.442297
                    50%
                                                  484.980155
                                                                                          3.009923
                                                                                                                        1.731530
                                                                                                                                                                     97.220408
                                                                                                                                                                                                                      0.803133
                    75%
                                                  502.420843
                                                                                         3.993998
                                                                                                                        2.370437
                                                                                                                                                                     98.379422
                                                                                                                                                                                                                      1.145900
                                                  519 994796
                                                                                          4 999955
                                                                                                                        2 999944
                                                                                                                                                                     99 499657
                                                                                                                                                                                                                      1 499440
                    max
 In [8]:
                  data.shape
 Out[8]:
                  (10000, 19)
 In [9]: data.size
 Out[9]: 190000
In [10]:
                 data.isnull().sum()
                  CastingTemperature(C)
                                                                                0
                                                                                0
                  RollingSpeed(m/min)
                   CoolingRate(C/s)
                                                                                0
                   ChemicalComposition(Al%)
                                                                                0
                   ChemicalComposition(Cu%)
                                                                                0
                   ChemicalComposition(Mg%)
                                                                                0
                   ChemicalComposition(Si%)
                                                                                0
                  ChemicalComposition(Fe%)
                                                                                0
                  CastBarEntryTemperature(C)
                                                                                0
                  EmulsionTemperature(C)
                                                                                0
                  AgingTemperature(C)
                                                                                0
                   EmulsionPressure(Bar)
                                                                                0
                  EmulsionConcentration(%)
                                                                                0
                  RodQuenchWaterPressure(Bar)
                                                                                0
                   FirstRollingTemperature(C)
                                                                                0
                   QuenchingTemperature(C)
                                                                                0
                  Elongation(%)
                                                                                0
                                                                                0
                   Conductivity(%IACS)
                  UTS(Mpa)
                                                                                0
                  dtype: int64
                  newdata = data.drop(["ChemicalComposition(Mg%)", "ChemicalComposition(Si%)", "CastBarEntryTemperature(C)", "Agia
                  newdata.head()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 19 columns):

Out[11]: ${\tt CastingTemperature(C)} \quad {\tt RollingSpeed(m/min)} \quad {\tt CoolingRate(C/s)} \quad {\tt ChemicalComposition(Al\%)} \quad {\tt ChemicalComposition(Cu\%)} \quad {\tt ChemicalCom$ 1.343437 0 517.738762 2.670609 2.124924 97.462870 1 516.073516 4.669137 1.951931 96.631344 0.341770 2 514.097405 4.114840 2.440798 97.577301 0.732053 3 458.507464 4.946501 1.770175 98.004922 1.253575 4 502.823147 3.941969 0.611059 99.038671 0.430471

In [12]: newdata.tail()

Out[12]: CastingTo

:		CastingTemperature(C)	RollingSpeed(m/min)	CoolingRate(C/s)	ChemicalComposition(AI%)	ChemicalComposition(Cu%)	Chemi
	9995	450.896112	1.591570	2.469588	95.141012	1.371089	
	9996	454.236645	2.620234	2.536631	99.230694	0.268305	
	9997	496.007223	2.725959	2.837583	97.023051	1.240611	
	9998	503.590094	1.449964	1.766031	96.396007	1.087624	
	9999	472.481583	4.524516	2.319007	99.419067	1.365180	

In [13]: newdata.sample(15)

Out[13]:

	CastingTemperature(C)	RollingSpeed(m/min)	CoolingRate(C/s)	ChemicalComposition(Al%)	ChemicalComposition(Cu%)	Chem
9198	463.832756	4.938848	1.319622	99.469287	0.795620	
95	474.497241	2.640713	0.949833	95.745777	0.270540	
9721	498.072065	2.435587	2.943607	98.882857	0.336209	
2469	509.676394	1.113680	1.313695	96.862746	1.191073	
9507	465.339521	3.621071	2.841663	96.286954	0.605596	
3388	509.577046	2.228630	1.371930	95.128693	1.370580	
1753	453.687235	3.485772	2.179596	95.365543	0.655866	
9564	498.764133	3.219259	1.292278	98.450415	0.957299	
9493	467.846824	4.170972	1.028801	98.643933	0.948184	
2436	504.257252	3.378568	1.061124	98.826781	0.228141	
4516	511.305888	3.389589	1.677753	97.438609	1.249815	
6629	475.738294	2.656471	2.395499	98.793396	1.152166	
5113	517.356043	2.191776	1.771043	96.918791	0.624277	
6208	478.434903	1.889639	1.909067	96.477892	0.992041	
3694	488.692848	3.365346	2.011290	98.938673	1.164052	

In [14]: newdata.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	CastingTemperature(C)	10000 non-null	float64
1	RollingSpeed(m/min)	10000 non-null	float64
2	CoolingRate(C/s)	10000 non-null	float64
3	ChemicalComposition(Al%)	10000 non-null	float64
4	ChemicalComposition(Cu%)	10000 non-null	float64
5	ChemicalComposition(Fe%)	10000 non-null	float64
6	<pre>EmulsionTemperature(C)</pre>	10000 non-null	float64
7	EmulsionPressure(Bar)	10000 non-null	float64
8	<pre>EmulsionConcentration(%)</pre>	10000 non-null	float64
9	Elongation(%)	10000 non-null	float64
10	Conductivity(%IACS)	10000 non-null	float64
11	UTS(Mpa)	10000 non-null	float64
1.	67 (64/40)		

dtypes: float64(12)
memory usage: 937.6 KB

```
Out[15]:
                                                       CastingTemperature(C) RollingSpeed(m/min) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComp
                                                                                                                                                    10000.000000
                                                                                                                                                                                                         10000.000000
                                                                                                                                                                                                                                                                                                                                                                                10000.000000
                                  count
                                                                                   10000.000000
                                                                                                                                                                                                                                                                                            10000.000000
                                                                                         485.100301
                                                                                                                                                                2.999822
                                                                                                                                                                                                                      1.741335
                                                                                                                                                                                                                                                                                                     97.236129
                                                                                                                                                                                                                                                                                                                                                                                             0.797193
                                  mean
                                                                                             20.200657
                                                                                                                                                                1.160946
                                                                                                                                                                                                                      0.724717
                                                                                                                                                                                                                                                                                                         1.302841
                                                                                                                                                                                                                                                                                                                                                                                             0.403958
                                         std
                                                                                                                                                                                                                                                                                                                                                                                             0.100020
                                       min
                                                                                         450.005874
                                                                                                                                                                 1.000230
                                                                                                                                                                                                                      0.500046
                                                                                                                                                                                                                                                                                                     95.000018
                                      25%
                                                                                                                                                                1.989422
                                                                                                                                                                                                                                                                                                                                                                                             0.442297
                                                                                         467.728565
                                                                                                                                                                                                                      1.111139
                                                                                                                                                                                                                                                                                                     96.099753
                                      50%
                                                                                         484.980155
                                                                                                                                                                3.009923
                                                                                                                                                                                                                      1.731530
                                                                                                                                                                                                                                                                                                     97.220408
                                                                                                                                                                                                                                                                                                                                                                                             0.803133
                                      75%
                                                                                          502.420843
                                                                                                                                                                3.993998
                                                                                                                                                                                                                      2.370437
                                                                                                                                                                                                                                                                                                      98.379422
                                                                                                                                                                                                                                                                                                                                                                                              1.145900
                                                                                                                                                                                                                                                                                                     99.499657
                                                                                                                                                                                                                                                                                                                                                                                              1.499440
                                      max
                                                                                          519.994796
                                                                                                                                                                 4.999955
                                                                                                                                                                                                                      2.999944
In [16]: newdata.shape
                                  (10000, 12)
Out[16]:
                                newdata.size
In [17]:
Out[17]:
                                  120000
In [18]: newdata.isnull().sum()
Out[18]:
                                 CastingTemperature(C)
                                                                                                                                     0
                                  RollingSpeed(m/min)
                                                                                                                                     0
                                  CoolingRate(C/s)
                                                                                                                                     0
                                  ChemicalComposition(Al%)
                                                                                                                                     0
                                  ChemicalComposition(Cu%)
                                                                                                                                     0
                                                                                                                                     0
                                  ChemicalComposition(Fe%)
                                  EmulsionTemperature(C)
                                                                                                                                     0
                                  EmulsionPressure(Bar)
                                                                                                                                     0
                                  EmulsionConcentration(%)
                                                                                                                                     0
                                                                                                                                     0
                                  Elongation(%)
                                  Conductivity(%IACS)
                                                                                                                                     0
                                                                                                                                     0
                                  UTS(Mpa)
                                  dtype: int64
 In [19]: X=newdata.drop(["CastingTemperature(C)", "RollingSpeed(m/min)", "CoolingRate(C/s)", "ChemicalComposition(Al%)",
                                 X.head()
                                          Elongation(%) Conductivity(%IACS)
Out[19]:
                                                                                                                                                            UTS(Mpa)
                                 0
                                                       79.004379
                                                                                                                       46.183274
                                                                                                                                                        151.816979
                                  1
                                                       78.148739
                                                                                                                       45.231306
                                                                                                                                                        149.161207
                                 2
                                                       76.589026
                                                                                                                       43.895326
                                                                                                                                                      141.849935
                                 3
                                                                                                                       43.427165
                                                       66.702291
                                                                                                                                                     157.580112
                                  4
                                                       73.947378
                                                                                                                       44.451037 163.040135
 In [20]: Y = newdata.drop(["Elongation(%)", "Conductivity(%IACS)", "UTS(Mpa)"], axis="columns")
Out[20]:
                                          CastingTemperature(C) RollingSpeed(m/min) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComp
                                 0
                                                                             517.738762
                                                                                                                                                                                                                                                                                         97.462870
                                                                                                                                                                                                                                                                                                                                                                                1.343437
                                                                                                                                                    2.670609
                                                                                                                                                                                                         2.124924
                                 1
                                                                             516.073516
                                                                                                                                                    4.669137
                                                                                                                                                                                                          1.951931
                                                                                                                                                                                                                                                                                         96.631344
                                                                                                                                                                                                                                                                                                                                                                                0.341770
                                 2
                                                                            514.097405
                                                                                                                                                    4.114840
                                                                                                                                                                                                         2.440798
                                                                                                                                                                                                                                                                                         97.577301
                                                                                                                                                                                                                                                                                                                                                                                0.732053
                                 3
                                                                             458.507464
                                                                                                                                                    4.946501
                                                                                                                                                                                                          1.770175
                                                                                                                                                                                                                                                                                         98.004922
                                                                                                                                                                                                                                                                                                                                                                                1.253575
                                  4
                                                                             502.823147
                                                                                                                                                    3.941969
                                                                                                                                                                                                         0.611059
                                                                                                                                                                                                                                                                                         99.038671
                                                                                                                                                                                                                                                                                                                                                                                0.430471
```

In [21]: X

Out[21]:		Elongation(%)	Conductivity(%IACS)	UTS(Mpa)
	0	79.004379	46.183274	151.816979
	1	78.148739	45.231306	149.161207
	2	76.589026	43.895326	141.849935
	3	66.702291	43.427165	157.580112
	4	73.947378	44.451037	163.040135
	9995	70.246691	42.916029	162.476887
	9996	70.575571	44.014963	151.391820
	9997	75.721106	45.213703	144.549297
	9998	77.175560	43.742988	154.972563
	9999	70.947047	44.780476	155.620165

10000 rows × 3 columns

Ιn	1001		,
ΤΠ	[22]	-	

Out[22]:		CastingTemperature(C)	RollingSpeed(m/min)	CoolingRate(C/s)	ChemicalComposition(Al%)	ChemicalComposition(Cu%)	Chemi
	0	517.738762	2.670609	2.124924	97.462870	1.343437	
	1	516.073516	4.669137	1.951931	96.631344	0.341770	
	2	514.097405	4.114840	2.440798	97.577301	0.732053	
	3	458.507464	4.946501	1.770175	98.004922	1.253575	
	4	502.823147	3.941969	0.611059	99.038671	0.430471	
	9995	450.896112	1.591570	2.469588	95.141012	1.371089	
	9996	454.236645	2.620234	2.536631	99.230694	0.268305	
	9997	496.007223	2.725959	2.837583	97.023051	1.240611	
	9998	503.590094	1.449964	1.766031	96.396007	1.087624	
	9999	472.481583	4.524516	2.319007	99.419067	1.365180	

10000 rows × 9 columns

In [24]: len(X_train)

Out[24]: 8000

In [25]: len(X_test)

Out[25]: 2000

In [26]: len(y_train)

Out[26]: 8000

In [27]: len(y_test)

Out[27]: 2000

In [28]: X_train.shape

Out[28]: (8000, 3)

In [29]: y_train.shape

Out[29]: (8000, 9)

In [30]: X_train.head()

```
Elongation(%) Conductivity(%IACS)
                                                                                  UTS(Mpa)
                                78 678840
                                                                45 336680 148 886003
                5595
                4969
                                78.022582
                                                                45.496135 163.030213
                                75.820067
                                                                44.676865
                                                                                149.717593
                  646
                4637
                                76.564725
                                                                46.666607
                                                                               161.883959
                6033
                                75 532052
                                                                44 950268 161 215067
In [31]: y train.head()
Out[31]:
                          CastingTemperature(C) RollingSpeed(m/min) CoolingRate(C/s) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComposition(Cu%) ChemicalComposition(Al%) ChemicalComposition(Cu%) ChemicalComp
                5595
                                           507.990703
                                                                              4.421008
                                                                                                         0.918457
                                                                                                                                                98.684930
                                                                                                                                                                                            0.674289
                4969
                                           507.463897
                                                                              4.482023
                                                                                                         1.238204
                                                                                                                                                 96.124499
                                                                                                                                                                                            1.282339
                                           510.047020
                                                                              4.008095
                                                                                                         1.358859
                                                                                                                                                 96.752980
                                                                                                                                                                                             1.070494
                  646
                4637
                                           504.664542
                                                                              2.624789
                                                                                                         2.536032
                                                                                                                                                 97.939341
                                                                                                                                                                                             1.444780
                6033
                                           513.483767
                                                                              2.795351
                                                                                                         0.968323
                                                                                                                                                97.202745
                                                                                                                                                                                             1.444487
In [32]:
                from sklearn.linear model import LinearRegression
                from sklearn.tree import DecisionTreeRegressor
                from sklearn.ensemble import RandomForestRegressor
In [33]: | lrmodel = LinearRegression()
                lrmodel.fit(X_train , y_train)
Out[33]: v LinearRegression
                LinearRegression()
In [34]: lrmodel.predict(X test)
                                                           2.97447795,
Out[34]: array([[491.1331561 ,
                                                                                    1.88232256, ..., 54.1774526 ,
                                  2.00655527, 12.6024731 ],
                             [465.03525745,
                                                           3.00480411.
                                                                                    1.4945368 , ...,
                                                                                                                  44.7935658 .
                                  2.00263861,
                                                        12.52027337],
                             [499.12196138,
                                                         2.99212901,
                                                                                   1.53683956, ...,
                                                                                                                   34.45437479.
                                  1.99608291, 12.50395742],
                             [492.58819874,
                                                           2.97117435,
                                                                                                                  37.09096122,
                                                                                    1.55849658, ...,
                                  1.9964039 ,
                                                         12.56408835],
                             [460.20120228,
                                                           2.99829521,
                                                                                    1.64756469, ...,
                                                                                                                   54.03467286,
                                  2.00763843,
                                                         12.5628714 ],
                             [487.90187171,
                                                           2.97470321,
                                                                                   1.62046985, ...,
                                                                                                                  42.02197512,
                                  1.9994544 ,
                                                        12.571433 ||)
In [35]: prediction = lrmodel.predict([[79.004379,46.183274,151.816979]])
                prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
                prediction
              C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
              alid feature names, but LinearRegression was fitted with feature names
                 warnings.warn(
Out[35]: [['513.10', '3.00', '2.11', '97.67', '0.79', '0.30', '58.98', '2.01', '12.53']]
In [37]: prediction = lrmodel.predict([[71.362238,44.513813,156.158610]])
                prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
                prediction_
              C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
              alid feature names, but LinearRegression was fitted with feature names
                 warnings.warn(
Out[37]: [['473.93', '2.97', '1.65', '97.13', '0.81', '0.30', '48.43', '2.00', '12.61']]
In [38]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
                X_train_prediction = lrmodel.predict(X_train)
                mae = mean_absolute_error(X_train_prediction , y_train)
                mse = mean_squared_error(X_train_prediction , y_train)
                rmse = np.sqrt(mse)
                r2 = r2_score(X_train_prediction , y_train)
                print(f'Mean Absolute Error (MAE): {mae}')
                print(f'Mean Squared Error (MSE): {mse}')
                print(f'Root Mean Squared Error (RMSE): {rmse}')
                print(f'R-squared (R2): {r2}')
```

```
Mean Squared Error (MSE): 17.639414466921213
             Root Mean Squared Error (RMSE): 4.199930293102638
             R-squared (R<sup>2</sup>): -2458.7523023089393
In [39]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
               X test prediction = lrmodel.predict(X test)
               mae = mean_absolute_error(X_test_prediction , y_test)
               mse = mean squared error(X test prediction , y test)
               rmse = np.sqrt(mse)
               r2 = r2_score(X_test_prediction , y_test)
               print(f'Mean Absolute Error (MAE): {mae}')
               print(f'Mean Squared Error (MSE): {mse}')
               print(f'Root Mean Squared Error (RMSE): {rmse}')
               print(f'R-squared (R2): {r2}')
             Mean Absolute Error (MAE): 2.256234704137137
             Mean Squared Error (MSE): 17.504045032625967
             Root Mean Squared Error (RMSE): 4.183783578607523
             R-squared (R<sup>2</sup>): -2457.3971791698496
In [40]: dtmodel = DecisionTreeRegressor()
               dtmodel.fit(X_train, y_train)
Out[40]: v DecisionTreeRegressor
               DecisionTreeRegressor()
In [41]: dtmodel.predict(X test)
Out[41]: array([[481.8876868 ,
                                                                              1.60216575, ..., 49.06833143,
                                                      3.54377527.
                                2.05310458, 10.04780125],
                           [453.3142597 ,
                                                     1.57371906,
                                                                             1.17953927, ..., 42.2691258,
                                2.96176095,
                                                    13.90828736],
                           [517.0555955 ,
                                                    1.74859173.
                                                                             1.12580022, ..., 42.36852553,
                               1.16751689, 12.98437956],
                           [503.5895542 ,
                                                      3.73659972,
                                                                              0.68876736, ..., 44.64712084,
                               1.10704769, 13.33995402],
                           [452.2030989 ,
                                                    2.6754889 ,
                                                                             0.73266691, ..., 58.01695
                               2.09349788, 12.02168878],
                           [491.2363573 ,
                                                     4.26556122,
                                                                             1.29533409, ..., 42.87313612,
                               1.18832813, 16.88855592]])
In [42]: prediction = dtmodel.predict([[79.004379,46.183274,151.816979]])
               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
               prediction
             C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
             alid feature names, but DecisionTreeRegressor was fitted with feature names
Out[42]: [['517.74', '2.67', '2.12', '97.46', '1.34', '0.49', '57.78', '2.02', '10.92']]
In [43]: prediction = dtmodel.predict([[71.362238,44.513813,156.158610]])
               prediction = [[f"{num:.2f}" for num in row] for row in prediction]
               prediction_
              \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have various and the packages of the packages 
             alid feature names, but DecisionTreeRegressor was fitted with feature names
               warnings.warn(
Out[43]: [['479.17', '1.18', '2.97', '96.47', '0.12', '0.43', '42.31', '2.49', '12.90']]
In [44]: from sklearn.metrics import mean absolute error, mean squared error, r2 score
               X_train_prediction = dtmodel.predict(X_train)
               mae = mean absolute error(X train prediction , y train)
               mse = mean_squared_error(X_train_prediction , y_train)
               rmse = np.sqrt(mse)
               r2 = r2_score(X_train_prediction , y_train)
               print(f'Mean Absolute Error (MAE): {mae}')
               print(f'Mean Squared Error (MSE): {mse}')
               print(f'Root Mean Squared Error (RMSE): {rmse}')
               print(f'R-squared (R2): {r2}')
             Mean Absolute Error (MAE): 0.0
             Mean Squared Error (MSE): 0.0
             Root Mean Squared Error (RMSE): 0.0
             R-squared (R^2): 1.0
In [45]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
               X_test_prediction = dtmodel.predict(X_test)
               mae = mean_absolute_error(X_test_prediction , y_test)
               mse = mean squared error(X test prediction , y test)
```

Mean Absolute Error (MAE): 2.268577305512877

```
rmse = np.sart(mse)
               r2 = r2_score(X_test_prediction , y_test)
               print(f'Mean Absolute Error (MAE): {mae}')
               print(f'Mean Squared Error (MSE): {mse}')
               print(f'Root Mean Squared Error (RMSE): {rmse}')
               print(f'R-squared (R2): {r2}')
             Mean Absolute Error (MAE): 3.0514889348682233
             Mean Squared Error (MSE): 33.9648430247367
             Root Mean Squared Error (RMSE): 5.827936429366462
             R-squared (R<sup>2</sup>): -0.6323788969970702
In [46]: rfmodel = RandomForestRegressor()
               rfmodel.fit(X train, y train)
Out[46]: ▼ RandomForestRegressor
               RandomForestRegressor()
In [47]: rfmodel.predict(X test)
Out[47]: array([[491.90796933,
                                                       3.10318342,
                                                                               1.906181 , ..., 55.61182774,
                                2.29092171, 11.21807569],
                            [458.04128724,
                                                      2.56207442,
                                                                               1.53305869, ..., 42.95525764,
                                2.26336186, 12.92943778],
                            [502.79098325, 3.24116577, 1.91782837, 11.89953071],
                                                                               1.48636661. .... 37.42069535.
                            [500.49878041,
                                                      3.29876936.
                                                                               1.08161794, ..., 40.04933959,
                                1.50842092, 13.33580036],
                                                                               1.44798921, ..., 58.06060792,
                            [457.27887991,
                                                      3.12238988.
                                2.02984274, 13.78695477],
                            [490.21672967,
                                                     2.943932
                                                                               1.43634694, ..., 40.52973099,
                                1.832185 , 13.20201121]])
In [48]: prediction = rfmodel.predict([[79.004379,46.183274,151.816979]])
               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
               prediction
              \verb|C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have various and the packages of the packages 
             alid feature names, but RandomForestRegressor was fitted with feature names
Out[48]: [['515.13', '3.15', '2.17', '97.27', '1.12', '0.44', '58.26', '2.15', '10.78']]
In [49]: prediction = rfmodel.predict([[71.362238,44.513813,156.158610]])
               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
               prediction
             C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
             alid feature names, but RandomForestRegressor was fitted with feature names
              warnings.warn(
Out[49]: [['468.60', '3.19', '1.74', '96.98', '0.62', '0.37', '48.53', '2.21', '12.85']]
In [50]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
               X_train_prediction = rfmodel.predict(X_train)
               mae = mean_absolute_error(X_train_prediction , y_train)
               mse = mean_squared_error(X_train_prediction , y_train)
               rmse = np.sqrt(mse)
               r2 = r2 score(X train_prediction , y_train)
               print(f'Mean Absolute Error (MAE): {mae}')
               print(f'Mean Squared Error (MSE): {mse}')
               print(f'Root Mean Squared Error (RMSE): {rmse}')
               print(f'R-squared (R2): {r2}')
             Mean Absolute Error (MAE): 0.8584005135892219
             Mean Squared Error (MSE): 2.644683253531503
             Root Mean Squared Error (RMSE): 1.6262482139979442
             R-squared (R<sup>2</sup>): 0.7243370454256639
In [51]: from sklearn.metrics import mean_absolute_error, mean squared error, r2 score
               X test prediction = rfmodel.predict(X test)
               mae = mean absolute error(X test prediction , y test)
               mse = mean squared error(X test prediction , y test)
               rmse = np.sqrt(mse)
               r2 = r2_score(X_test_prediction , y_test)
               print(f'Mean Absolute Error (MAE): {mae}')
               print(f'Mean Squared Error (MSE): {mse}')
               print(f'Root Mean Squared Error (RMSE): {rmse}')
               print(f'R-squared (R2): {r2}')
```

```
Mean Squared Error (MSE): 18.65789484129651
                           Root Mean Squared Error (RMSE): 4.319478538121993
                           R-squared (R<sup>2</sup>): -6.339314479471753
In [52]: input data = (71.362238,44.513813,156.158610)
                                input_data_as_numpy_array = np.asarray(input_data)
                               input_data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
                               prediction = lrmodel.predict(input_data_reshaped)
                               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
                               prediction
                            \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have various and the packages of the packages 
                           alid feature names, but LinearRegression was fitted with feature names
Out[52]: [['473.93', '2.97', '1.65', '97.13', '0.81', '0.30', '48.43', '2.00', '12.61']]
In [53]: input_data = (71.362238,44.513813,156.158610)
                               input data as numpy array = np.asarray(input data)
                               input data reshaped = input data as numpy array.reshape(1 , -1)
                               prediction = dtmodel.predict(input data reshaped)
                               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
                               prediction
                            \verb| C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have volume to the packages of the packages o
                           alid feature names, but DecisionTreeRegressor was fitted with feature names
                                warnings.warn(
Out[53]: [['479.17', '1.18', '2.97', '96.47', '0.12', '0.43', '42.31', '2.49', '12.90']]
In [54]: input_data = (66.702,43.465,157.58)
                               input_data_as_numpy_array = np.asarray(input_data)
                               input_data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
                               prediction = lrmodel.predict(input_data_reshaped)
                               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
                               prediction
                            \verb|C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have value of the packages of 
                           alid feature names, but LinearRegression was fitted with feature names
Out[54]: [['450.09', '2.95', '1.37', '96.80', '0.81', '0.30', '41.78', '2.00', '12.64']]
In [55]: Elongation = float(input(""))
                               Conductivity = float(input(""))
                               UTS = float(input(""))
In [56]: input data = (Elongation, Conductivity, UTS)
                               input data as numpy array = np.asarray(input data)
                               input data reshaped = input data as numpy array.reshape(1 , -1)
                               prediction = lrmodel.predict(input_data_reshaped)
                               prediction_ = [[f"{num:.2f}" for num in row] for row in prediction]
                               prediction
                            \verb| C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have volume to the packages of the packages o
                           alid feature names, but LinearRegression was fitted with feature names
                                warnings.warn(
Out[56]: [['450.09', '2.95', '1.37', '96.80', '0.81', '0.30', '41.78', '2.00', '12.64']]
In [57]: import pickle
                               with open("aluminium parameter model pickle" , "wb") as file:
                                                pickle.dump(lrmodel , file)
In [58]: import numpy as np
                               import pickle
                               with open("aluminium parameter model pickle" , "rb") as file:
                                                lrmodel = pickle.load(file)
In [59]: input_data = (66.702,43.465,157.58)
                               input_data_as_numpy_array = np.asarray(input_data)
                               input_data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
                               prediction = lrmodel.predict(input_data_reshaped)
                               prediction = [[f"{num:.2f}" for num in row] for row in prediction]
                               prediction
                           C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
                           alid feature names, but LinearRegression was fitted with feature names
                                 warnings.warn(
Out[59]: [['450.09', '2.95', '1.37', '96.80', '0.81', '0.30', '41.78', '2.00', '12.64']]
```

Mean Absolute Error (MAE): 2.3091897142144626

In []:

