with Python

NumPy Pandas

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1.1

```
Python
NumPy -
Pandas -
Matplotlib/Seaborn -
Scikit-learn -
```

```
#
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

# 1.
# 2.
# 3. EDA
# 4.
# 5.
```

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1.3 NumPy -

```
import numpy as np
arr1 = np.array([1, 2, 3, 4, 5])
arr2 = np.array([[1, 2, 3], [4, 5, 6]])
print(f"1 : {arr1}")
                               # [1 2 3 4 5]
print(f"2 :\n{arr2}")
                               # [[1 2 3], [4 5 6]]
print(f" : {arr2.shape}")
                                  # (2, 3)
print(f" : {arr1.dtype}") # int64
zeros = np.zeros((3, 4))
                                    # 1
ones = np.ones((2, 3))
                                    # [0 2 4 6 8]
range_arr = np.arange(0, 10, 2)
linspace = np.linspace(0, 1, 5)
                                    # [0. 0.25 0.5 0.75 1. ]
random_arr = np.random.rand(3, 3)
arr = np.array([1, 2, 3, 4, 5])
print(arr * 2)
                                     # [2 4 6 8 10]
                                     # [1 4 9 16 25]
print(arr ** 2)
                                    # [1. 1.41 1.73 2. 2.24]
print(np.sqrt(arr))
```

1.4 NumPy

```
#
data = np.random.randn(1000) # 1000

#
print(f" : {np.mean(data):.3f}")
print(f" : {np.std(data):.3f}")
```

```
print(f" : {np.min(data):.3f}")
print(f" : {np.max(data):.3f}")
print(f" : {np.median(data):.3f}")
matrix = np.arange(12).reshape(3, 4)
print(f" : {matrix.shape}")
print(f" : {matrix.T.shape}")
print(f" : {matrix[0]}")
print(f" : {matrix[:, -1]}")
print(f" : {matrix[matrix > 5]}")
arr1 = np.array([1, 2, 3])
arr2 = np.array([4, 5, 6])
combined = np.concatenate([arr1, arr2]) # [1 2 3 4 5 6]
stacked = np.stack([arr1, arr2]) # [[1 2 3], [4 5 6]]
A = np.array([[1, 2], [3, 4]])
B = np.array([[5, 6], [7, 8]])
print(f" : \n{np.dot(A, B)}")
print(f" : \n{np.linalg.inv(A)}")
print(f" : {np.linalg.eigvals(A)}")
```

1.5 Pandas -

```
import pandas as pd

# Series 1
series = pd.Series([1, 3, 5, 7, 9], index=['a', 'b', 'c', 'd', 'e'])
print(series)

# DataFrame 2
data = {
    ' ': [' ', ' ', ' ', ' '],
    ' ': [25, 30, 35, 28],
```

```
'': [''', ''', '''],
'': [5000000, 4500000, 7000000, 6000000]
}

df = pd.DataFrame(data)
print(df)
```

```
    0
    25
    5000000

    1
    30
    4500000

    2
    35
    7000000

    3
    28
    6000000
```

1.6 Pandas

```
print(df.info())
print(df.describe())
print(df.head())
                                  # 5
print(df.tail(3))
                                  # 3
print(df[' '])
print(df[[' ', ' ']])
print(df.iloc[0])
print(df.loc[df[' '] > 30])
  = df[df[' '] > 5000000]
   = df[df[' '] < 30]
   = df[df[' '].str.contains(' ')]
  = df.sort_values(' ')
   = df.sort_values(' ', ascending=False)
   = df.groupby(' ').agg({
   ' ': 'mean',
```

```
# CSV
# df = pd.read_csv('data.csv', encoding='utf-8')
# df = pd.read_excel('data.xlsx')
# df = pd.read_json('data.json')
# df = pd.read_sql('SELECT * FROM table', connection)
import pandas as pd
import numpy as np
np.random.seed(42)
dates = pd.date_range('2023-01-01', '2023-12-31', freq='D')
products = [' A', ' B', ' C', ' D']
sales_data = []
for date in dates:
    for product in products:
        sales = np.random.poisson(lam=50) + np.random.randint(0, 100)
        price = np.random.uniform(1000, 5000)
        sales_data.append({
            ' ': date,
            ' ': product,
            ' ': sales,
            ' ': round(price),
            ' ': sales * round(price)
        })
```

```
sales_df = pd.DataFrame(sales_data)

#
sales_df.to_csv('sales_data.csv', index=False, encoding='utf-8')
print(" ")
print(sales_df.head())
```

1.8 - Matplotlib

```
import matplotlib.pyplot as plt
plt.style.use('default') #
plt.rcParams['font.family'] = 'DejaVu Sans' #
fig, axes = plt.subplots(2, 2, figsize=(12, 8))
x = np.linspace(0, 10, 100)
y = np.sin(x)
axes[0, 0].plot(x, y, 'b-', linewidth=2)
axes[0, 0].set_title('Sine Wave')
axes[0, 0].grid(True)
x_scatter = np.random.randn(100)
y_scatter = 2 * x_scatter + np.random.randn(100)
axes[0, 1].scatter(x_scatter, y_scatter, alpha=0.6)
axes[0, 1].set_title('Scatter Plot')
data_hist = np.random.normal(0, 1, 1000)
axes[1, 0].hist(data_hist, bins=30, alpha=0.7, color='green')
axes[1, 0].set_title('Histogram')
categories = ['A', 'B', 'C', 'D']
values = [23, 17, 35, 29]
axes[1, 1].bar(categories, values, color='orange')
axes[1, 1].set_title('Bar Chart')
```

```
plt.tight_layout()
plt.show()

#

plt.figure(figsize=(10, 6))

#

months = ['1', '2', '3', '4', '5', '6']
product_a = [120, 135, 158, 142, 167, 183]
product_b = [98, 112, 126, 139, 145, 156]

plt.plot(months, product_a, 'o-', label=' A', linewidth=2, markersize=8)
plt.plot(months, product_b, 's-', label=' B', linewidth=2, markersize=8)
plt.xlabel(' ')
plt.ylabel(' ')
plt.title(' ')
plt.title(' ')
plt.legend()
plt.grid(True, alpha=0.3)
plt.show()
```

```
#
def analyze_sales_data():
    #
    df = sales_df.copy()

#    datetime
df[' '] = pd.to_datetime(df[' '])
df[' '] = df[' '].dt.month
df[' '] = df[' '].dt.day_name()

#
    print("=== ===")
    print(f" : {df[' '].sum():,} ")
    print(f" : {df.groupby(' ')[' '].sum().mean():,.0f} ")
    print(f" : {df.groupby(' ')[' '].sum().max():,} ")

#
```

```
print("\n===
               ===")
product_analysis = df.groupby(' ').agg({
    ' ': ['sum', 'mean'],
    ' ': 'sum',
    ' ': 'mean'
}).round(0)
print(product_analysis)
monthly_sales = df.groupby(' ')[' '].sum()
weekday_sales = df.groupby(' ')[' '].sum()
correlation = df[[' ', ' ', ' ']].corr()
print("\n===
                ===")
print(correlation)
fig, axes = plt.subplots(2, 2, figsize=(15, 10))
product_total = df.groupby(' ')[' '].sum()
axes[0, 0].pie(product_total.values, labels=product_total.index, autopct='%1.1f%%')
axes[0, 0].set_title('
                       ')
axes[0, 1].plot(monthly_sales.index, monthly_sales.values, 'o-', linewidth=2)
axes[0, 1].set_title(' ')
axes[0, 1].set_xlabel(' ')
axes[0, 1].set_ylabel(' ')
axes[0, 1].grid(True)
# vs
axes[1, 0].scatter(df[' '], df[' '], alpha=0.6)
axes[1, 0].set_title(' vs ')
axes[1, 0].set_xlabel(' ')
axes[1, 0].set_ylabel(' ')
weekday_order = ['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Su
```

```
weekday_sales_ordered = weekday_sales.reindex(weekday_order)
axes[1, 1].bar(range(7), weekday_sales_ordered.values)
axes[1, 1].set_title(' ')
axes[1, 1].set_xlabel(' ')
axes[1, 1].set_ylabel(' ')
axes[1, 1].set_xticks(range(7))
axes[1, 1].set_xticklabels([' ', ' ', ' ', ' ', ' ', ' ', ' '])

plt.tight_layout()
plt.show()

return df

#
analyzed_df = analyze_sales_data()
```

1.10 - Scikit-learn

```
# DataFrame
housing_data = pd.DataFrame({
        ' ': area,
        ' ': station_distance,
        ' ': rooms,
        ' ': price
})
print(" :")
print(housing_data.head())
print(f"\n : {housing_data.shape}")
print("\n :")
print(housing_data.describe())
```

```
#
X = housing_data[[' ', ' ', ' ', ' ']]
y = housing_data[' ']

#
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)

print(f" : {X_train.shape[0]} ")
print(f" : {X_test.shape[0]} ")

# 1:
lr_model = LinearRegression()
lr_model.fit(X_train, y_train)

# 2:
rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)

#
lr_pred = lr_model.predict(X_test)
rf_pred = rf_model.predict(X_test)
```

```
def evaluate_model(y_true, y_pred, model_name):
   mse = mean_squared_error(y_true, y_pred)
    rmse = np.sqrt(mse)
   r2 = r2_score(y_true, y_pred)
   print(f"\n=== {model name} ===")
   print(f"RMSE: {rmse:.2f}")
   print(f"R2 : {r2:.4f}")
   return rmse, r2
lr_rmse, lr_r2 = evaluate_model(y_test, lr_pred, " ")
rf_rmse, rf_r2 = evaluate_model(y_test, rf_pred, "
feature_importance = pd.DataFrame({
   ' ': X.columns,
    ' ': rf_model.feature_importances_
}).sort_values(' ', ascending=False)
print("\n
            :")
print(feature_importance)
fig, axes = plt.subplots(1, 3, figsize=(18, 5))
axes[0].scatter(y_test, lr_pred, alpha=0.6)
axes[0].plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--', lw=2)
axes[0].set_xlabel('
                      ')
axes[0].set_ylabel(' ')
axes[0].set_title(f' (R^2 = \{lr_r2:.3f\})')
  VS
axes[1].scatter(y_test, rf_pred, alpha=0.6)
axes[1].plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--', lw=2)
axes[1].set_xlabel('
                      ')
axes[1].set_ylabel('
axes[1].set_title(f' (R^2 = \{rf_r2:.3f\})')
axes[2].barh(feature_importance[' '], feature_importance[' '])
```

```
axes[2].set_xlabel(' ')
axes[2].set_title('
plt.tight_layout()
plt.show()
new_house = pd.DataFrame({
   ' ': [80],
    ' ': [5],
    ' ': [8],
    ' ': [3]
})
lr_prediction = lr_model.predict(new_house)[0]
rf_prediction = rf_model.predict(new_house)[0]
print(f"\n
print(f" : {new_house[' '].iloc[0]}, : {new_house[' '].iloc[0]}, : {new_house[' '].iloc
print(f" : {lr_prediction:,.0f} ")
           : {rf_prediction:,.0f} ")
print(f"
```

1.13 :

```
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plt

#
np.random.seed(42)
n_customers = 500

#
age = np.random.normal(40, 12, n_customers)
income = np.random.normal(500, 150, n_customers) #
spending = np.random.normal(300, 100, n_customers) #
frequency = np.random.poisson(12, n_customers) #
```

```
age = np.clip(age, 18, 80)
income = np.clip(income, 200, 1000)
spending = np.clip(spending, 50, 800)
customer_data = pd.DataFrame({
   ' ': age,
   ' ': income,
   ' ': spending,
   ' ': frequency
})
print("
          :")
print(customer_data.head())
print("\n :")
print(customer_data.describe())
scaler = StandardScaler()
scaled_data = scaler.fit_transform(customer_data)
# K-means
kmeans = KMeans(n_clusters=4, random_state=42)
clusters = kmeans.fit_predict(scaled_data)
customer_data[' '] = clusters
cluster_analysis = customer_data.groupby(' ').agg({
   ' ': 'mean',
   ' ': 'mean',
    ' ': 'mean',
   ' ': 'mean'
}).round(1)
print("\n :")
print(cluster_analysis)
cluster_labels = {
0: " ",
```

```
1: " ",
   2: " ",
    3: " "
}
customer_data[' '] = customer_data[' '].map(cluster_labels)
fig, axes = plt.subplots(2, 2, figsize=(15, 12))
# vs
scatter = axes[0, 0].scatter(customer_data[' '], customer_data[' '],
                          c=customer_data[' '], cmap='viridis', alpha=0.7)
axes[0, 0].set_xlabel('
axes[0, 0].set_ylabel('
                       ')
axes[0, 0].set_title(' vs ')
plt.colorbar(scatter, ax=axes[0, 0])
  VS
scatter2 = axes[0, 1].scatter(customer_data[' '], customer_data[' '],
                           c=customer_data[' '], cmap='viridis', alpha=0.7)
axes[0, 1].set_xlabel(' ')
axes[0, 1].set_ylabel(' ')
axes[0, 1].set_title(' vs ')
cluster_counts = customer_data[' '].value_counts()
axes[1, 0].pie(cluster_counts.values, labels=cluster_counts.index, autopct='%1.1f%%')
axes[1, 0].set_title('
                        ')
cluster_means = customer_data.groupby(' ')[[' ', ' ']].mean()
cluster_means.plot(kind='bar', ax=axes[1, 1])
axes[1, 1].set_title('
axes[1, 1].set_ylabel('
axes[1, 1].tick_params(axis='x', rotation=45)
plt.tight_layout()
plt.show()
print("\n=== ===")
```

```
for cluster_id, label in cluster_labels.items():
    cluster_data = customer_data[customer_data[' '] == cluster_id]
    size = len(cluster_data)
    avg_income = cluster_data[' '].mean()
    avg_spending = cluster_data[' '].mean()
    avg_frequency = cluster_data[' '].mean()
    print(f"\n{label} ({size}, {size/len(customer_data)*100:.1f}%)")
   print(f" : {avg_income:.0f} ")
print(f" : {avg_spending:.0f} ")
    print(f" : {avg_frequency:.1f} ")
    if cluster_id == 0:
                              ")
        print(" :
    elif cluster_id == 1:
        print(" : VIP ")
    elif cluster_id == 2:
        print(" :
                               ")
    else:
                            ")
        print(" :
```

- 1. **NumPy** -
- 2. Pandas -
- 3. Matplotlib -
- 4. Scikit-learn -
- 5.

- Seaborn -
- Plotly -
- TensorFlow/PyTorch -
- Jupyter Notebook -
- - Kaggle

1.16.1

: - NumPy - Pandas - Scikit-learn - Kaggle -

1.17

:

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