```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

data = pd.read_csv('/content/mission5.csv')
```

data.head()

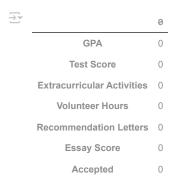
$\overrightarrow{\exists}$		GPA	Test Score	Extracurricular Activities	Volunteer Hours	Recommendation Letters	Essay Score	Accepted
	0	2.75	1374	5	16	2	9	No
	1	3.90	1542	7	27	3	3	Yes
	2	3.46	1040	4	29	1	9	No
	3	3.20	1363	3	28	2	1	No
	4	2.31	895	1	5	1	1	No

data['Accepted'] = data['Accepted'].map({"No":0, "Yes":1})

### data.head()

$\overline{\Rightarrow}$		GPA	Test Score	Extracurricular Activities	Volunteer Hours	Recommendation Letters	Essay Score	Accepted
	0	2.75	1374	5	16	2	9	0
	1	3.90	1542	7	27	3	3	1
	2	3.46	1040	4	29	1	9	0
	3	3.20	1363	3	28	2	1	0
	4	2.31	895	1	5	1	1	0

data.isna().sum()



dtype: int64

from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
data['Accepted'] = le.fit\_transform(data['Accepted'])

# data.head()

<del>-</del>		GPA	Test Score	Extracurricular Activities	Volunteer Hours	Recommendation Letters	Essay Score	Accepted
	0	2.75	1374	5	16	2	9	0
	1	3.90	1542	7	27	3	3	1
	2	3.46	1040	4	29	1	9	0
	3	3.20	1363	3	28	2	1	0
	4	2.31	895	1	5	1	1	0
	4							

from sklearn.preprocessing import StandardScaler

```
scaler = StandardScaler()
x = data.drop('Accepted', axis=1)
x = scaler.fit_transform(x)
F [[-0.32857578 0.57268455 0.41764096 -0.40939893 0.19576351 1.15296752]
      1.63284605 1.33301014 1.15849971 0.54162715 1.44375591 -0.94084887
      0.882389 -0.93891514 0.04721159 0.71454099 -1.05222889 1.15296752]
      0.43893711  0.52290132  -0.32321779  0.62808407  0.19576351  -1.63878766
     [-1.07903283 -1.59514854 -1.06407654 -1.36042501 -1.05222889 -1.63878766]
     [-1.07903283 1.29227841 0.41764096 1.14682557 -1.05222889 -0.59187947]
     -1.40309382 0.16536727 0.41764096 1.3197394 0.19576351 1.15296752
     1.34289674 -0.18764104 -1.43450591 0.19579949 -1.05222889 0.10605933]
     0.43893711 -0.98417262 1.52892908 0.62808407 1.44375591 -0.94084887
     0.81416563 1.36016462 0.41764096 0.8009979 -1.05222889 -1.63878766
     1.70106942 -1.25571747 -0.32321779 0.97391174 0.19576351 1.15296752
     [-0.89141857 -1.86669339 -0.69364716 0.88745482 -1.05222889 1.15296752]
     -0.99375362  0.34639717  1.89935846  0.10934257  0.19576351  0.45502872
     [-0.97669778 -1.30550069 -0.69364716 0.97391174 0.19576351 -0.59187947
    [-0.56735757 1.13387725 -0.69364716 -1.61979576 -1.05222889 -0.94084887]
     0.18309948 -0.92986365 -0.32321779 -0.32294201 1.44375591 1.50193692]
    -
[-0.14096152 -1.64040601 0.78807033 0.2822564 -1.05222889 -0.24291007]
     [-0.61852509 1.57287476 -0.32321779 0.8009979 0.19576351 -0.24291007]
      0.47304879  0.58626179  1.52892908  -1.61979576  0.19576351  -0.94084887
     -1.13020035 1.05693954 -1.43450591 1.57911015 -1.05222889 1.15296752]
     1.07000326 1.57287476 1.15849971 0.10934257 0.19576351 -0.24291007
     [-0.92553025 -1.90289937 -1.43450591 -1.70625268 -1.05222889 0.45502872]
     -1.45426135 -0.49086613 -1.06407654 0.36871332 0.19576351 -0.24291007
     -1.0278653 -0.53159786 1.89935846 1.57911015 1.44375591 1.50193692]
     [-1.38603798  0.86685814  -0.69364716  0.97391174  0.19576351  0.45502872]
     1.68401358 -0.45466015 1.89935846 1.49265324 -1.05222889 -0.24291007
     1.15528247 0.1110583 1.52892908 0.62808407 -1.05222889 -0.94084887]
     [-0.56735757 -0.98417262 -0.32321779 0.36871332 1.44375591 0.45502872]
     -1.26664709 -1.16972827 -1.43450591 1.14682557 0.19576351 -1.28981826]
     0.72888642 -1.0113271 -1.06407654 0.2822564 -1.05222889 1.15296752
     -0.10684983 -0.28720749 -1.43450591 0.19579949 1.44375591 1.50193692
     -1.19842372 -0.32341347 0.04721159 -0.75522659 0.19576351 1.50193692
     [ 0.08076443 -0.74883374  0.04721159 -1.27396809  1.44375591 -1.63878766]
     [-1.48837303 0.83517791 0.78807033 1.23328249 -1.05222889 0.10605933]
     1.49639931 0.81254917 1.52892908 -0.15002818 -1.05222889 0.45502872
     0.66066306 -0.85745168 -0.69364716 -1.18751118 1.44375591 1.50193692
     [-0.55030173 -0.40487693 -0.69364716 -0.49585584 1.44375591 1.15296752]
     0.16604364 1.18366047 -0.69364716 -0.66876968 -1.05222889 -1.28981826]
      [-0.97669778 -0.16953805 1.15849971 0.10934257 -1.05222889 -1.28981826]
     1.70106942 0.90758987 0.41764096 -0.58231276 -1.05222889 -0.24291007]
      1.03589158 -0.59043258 1.15849971 0.54162715 0.19576351 -0.24291007]
      1.59873437 1.0343108 -1.43450591 1.06036865 1.44375591 0.10605933]
      1.44523179 -1.0113271 1.15849971 -1.70625268 -1.05222889 -0.94084887
     0.43893711 -0.97059537 -0.32321779 0.88745482 -1.05222889 0.80399812
     -1.30075877 0.97547608 1.15849971 0.02288565 0.19576351 0.10605933]
    [-0.94258609 -0.33699071 -0.32321779 0.2822564 0.19576351 -0.59187947]
```

data.describe()

7		GPA	Test Score	Extracurricular Activities	Volunteer Hours	Recommendation Letters	Essay Score	Accepted
	count	102.000000	102.000000	102.000000	102.000000	102.000000	102.000000	102.000000
	mean	2.942647	1247.460784	3.872549	20.735294	1.843137	5.696078	0.500000
	std	0.589205	222.049128	2.712901	11.623573	0.805244	2.879732	0.502469
	min	2.010000	801.000000	0.000000	0.000000	1.000000	1.000000	0.000000
	25%	2.392500	1040.500000	2.000000	12.000000	1.000000	3.000000	0.000000
	50%	2.965000	1263.000000	3.000000	22.000000	2.000000	5.500000	0.500000
	75%	3.460000	1444.000000	6.000000	30.750000	2.750000	9.000000	1.000000
	max	3.970000	1597.000000	9.000000	39.000000	3.000000	10.000000	1.000000
	4							

```
plt.figure(figsize=(10,10))
sns.histplot(x=data['GPA'], color="red", kde=True)
plt.xlabel('GPA')
plt.ylabel('Count')
plt.title('Distribution of GPA')
plt.show()
```

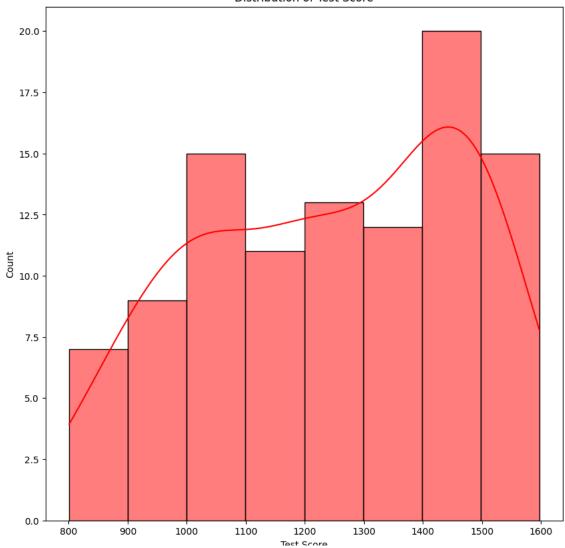


# Distribution of GPA 16 14 12 10 8 6 4 2 2.00 2.25 3.75 2.50 2.75 3.00 3.25 3.50 4.00 GPA

```
plt.figure(figsize=(10,10))
sns.histplot(x=data['Test Score'], color="red", kde=True)
plt.xlabel('Test Score')
plt.ylabel('Count')
plt.title('Distribution of Test Score')
plt.show()
```

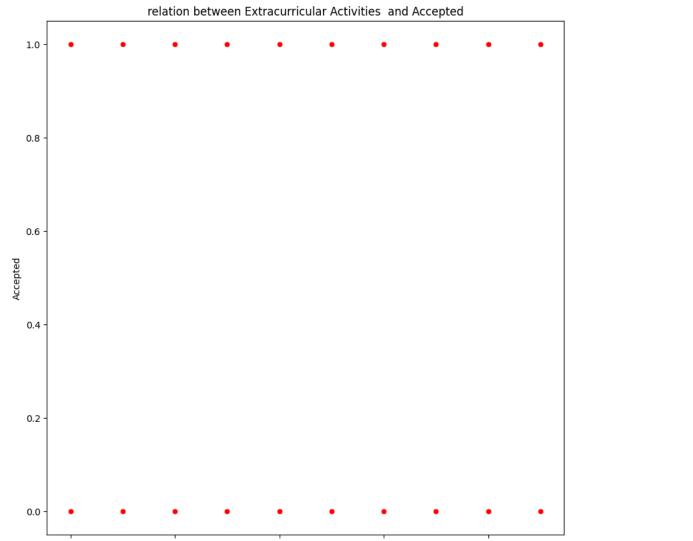
₹

## Distribution of Test Score



plt.figure(figsize=(10,10))
sns.scatterplot(x=data['Extracurricular Activities'], y=data['Accepted'], color="red")
plt.xlabel('Extracurricular Activities')
plt.ylabel('Accepted')
plt.title('relation between Extracurricular Activities and Accepted')
plt.show()



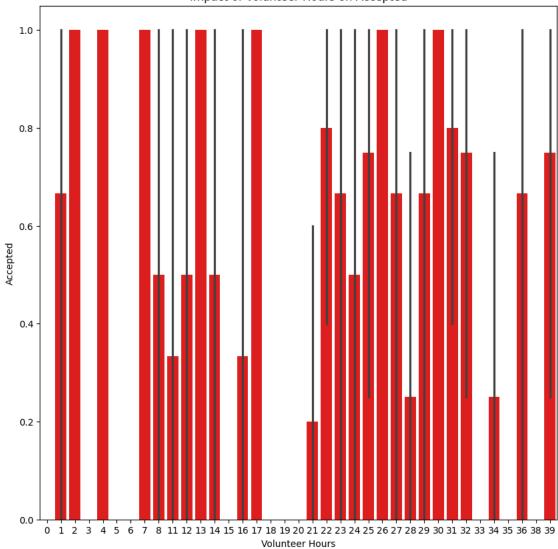


Extracurricular Activities

```
plt.figure(figsize=(10,10))
sns.barplot(x=data['Volunteer Hours'], y=data['Accepted'], color="red")
plt.xlabel('Volunteer Hours')
plt.ylabel('Accepted')
plt.title('Impact of Volunteer Hours on Accepted')
plt.show()
```



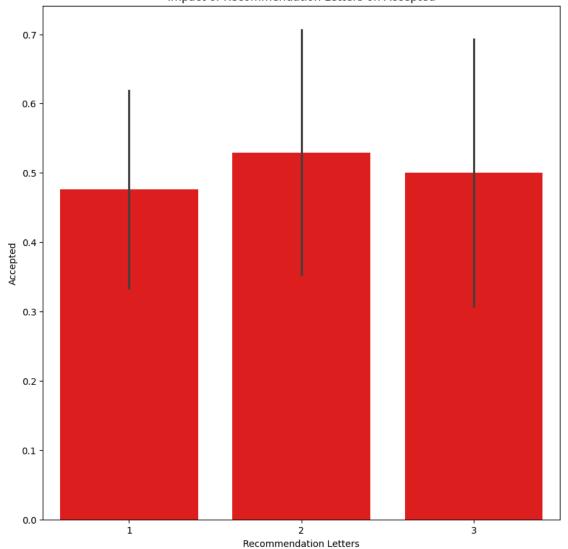




plt.figure(figsize=(10,10))
sns.barplot(x=data['Recommendation Letters'], y=data['Accepted'], color="red")
plt.xlabel('Recommendation Letters')
plt.ylabel('Accepted')
plt.title('Impact of Recommendation Letters on Accepted')
plt.show()



### Impact of Recommendation Letters on Accepted



```
array([[-0.32857578, 0.57268455, 0.41764096, -0.40939893, 0.19576351,
         1.15296752],
       [ 1.63284605, 1.33301014, 1.15849971, 0.54162715, 1.44375591,
        -0.94084887],
                   , -0.93891514, 0.04721159, 0.71454099, -1.05222889,
       [ 0.882389
         1.15296752],
       [ 0.43893711, 0.52290132, -0.32321779, 0.62808407, 0.19576351,
        -1.63878766],
       [-1.07903283, -1.59514854, -1.06407654, -1.36042501, -1.05222889,
        -1.63878766],
       [-1.07903283, 1.29227841, 0.41764096, 1.14682557, -1.05222889,
        -0.59187947],
       [-1.40309382, 0.16536727, 0.41764096, 1.3197394, 0.19576351,
         1.15296752],
       [ 1.34289674, -0.18764104, -1.43450591, 0.19579949, -1.05222889,
         0.10605933],
       [ 0.43893711, -0.98417262, 1.52892908, 0.62808407, 1.44375591, 
        -0.94084887],
       [ 0.81416563, 1.36016462, 0.41764096, 0.8009979 , -1.05222889,
        -1.63878766],
       [-1.53954056, 0.93474435, -0.69364716, 1.14682557, 0.19576351,
        -0.59187947],
       [ 1.70106942, -1.25571747, -0.32321779, 0.97391174, 0.19576351,
         1.15296752],
       [ 1.22350584, 0.41880913, -0.32321779, -0.06357126, 0.19576351,
        -0.94084887],
       [-0.89141857, -1.86669339, -0.69364716, 0.88745482, -1.05222889,
         1.15296752],
       [-0.99375362, 0.34639717, 1.89935846, 0.10934257, 0.19576351,
         0.45502872],
       [-0.97669778, -1.30550069, -0.69364716, 0.97391174, 0.19576351,
        -0.59187947],
```

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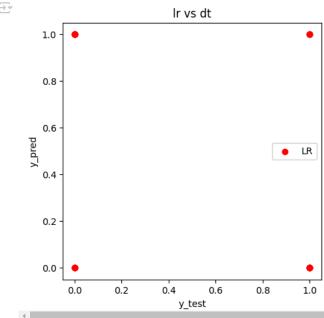
97 98 99

100 101

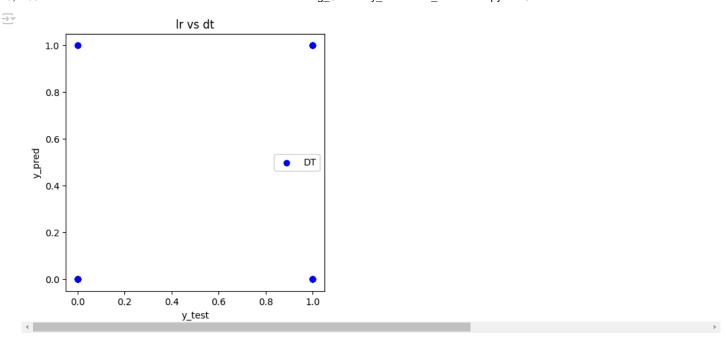
```
[-0.56735757, 1.13387725, -0.69364716, -1.61979576, -1.05222889,
             -0.940848871,
            [ 0.18309948, -0.92986365, -0.32321779, -0.32294201, 1.44375591,
              1.50193692],
            [-0.14096152, -1.64040601, 0.78807033, 0.2822564, -1.05222889,
             -0.24291007],
            [-0.61852509, 1.57287476, -0.32321779, 0.8009979, 0.19576351,
             -0.24291007],
            [\ 0.47304879,\ 0.58626179,\ 1.52892908,\ -1.61979576,\ 0.19576351,
              -0.94084887],
            [-1.13020035, 1.05693954, -1.43450591, 1.57911015, -1.05222889,
              1.15296752],
            [-0.61852509, 0.49122109, 1.15849971, 0.19579949, 1.44375591,
             -0.59187947],
            [-0.36268746, 0.5681588 , 0.78807033, 0.88745482, 0.19576351,
             -0.24291007],
            [-0.05568231, 0.89401263, -1.06407654, 0.02288565, 0.19576351,
             -0.59187947],
            [ 1.07000326, 1.57287476, 1.15849971, 0.10934257, 0.19576351,
             -0.24291007],
            [-0.92553025, -1.90289937, -1.43450591, -1.70625268, -1.05222889,
              0.45502872],
            [ \ 0.1489878 \ , \ 0.77634319, \ 1.52892908, \ 0.45517024, \ 0.19576351,
              1.15296752],
            [ 0.40482543, 0.48669534, 1.52892908, -1.70625268, 1.44375591,
y = data['Accepted']
           Accepted
                  0
                  0
                  \cap
                  \cap
                  0
     102 rows × 1 columns
     dtuna: int64
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
lr = LogisticRegression()
lr.fit(x_train, y_train)
dt = DecisionTreeClassifier()
dt.fit(x_train, y_train)
y_pred_lr = lr.predict(x_test)
y_pred_dt = dt.predict(x_test)
```

print("Accuracy score for lr:", accuracy\_score(y\_test, y\_pred\_lr)) print("Accuracy score for dt:", accuracy\_score(y\_test, y\_pred\_dt))

```
1/17/25, 2:33 PM
    Accuracy score for 1r: 0.38095238095238093
         Accuracy score for dt: 0.6190476190476191
    print("classification report for lr:", classification_report(y_test, y_pred_lr))
    print("classification report for dt:", classification_report(y_test, y_pred_dt))
    classification report for 1r:
                                                      precision
                                                                   recall f1-score support
                                                 0.38
                    0
                            9.49
                                       0.36
                                                             11
                            0.36
                                       0.40
                                                 0.38
                                                             10
             accuracy
                                                 0.38
                                                             21
            macro avg
                                       0.38
                            0.38
                                                 0.38
                                                             21
         weighted avg
                            0.38
                                       0.38
                                                 0.38
                                                             21
         classification report for dt:
                                                      precision
                                                                   recall f1-score support
                                       0.73
                                                 0.67
                    0
                            0.62
                                                             11
                    1
                            0.62
                                      0.50
                                                 0.56
                                                             10
                                                             21
                                                 0.62
             accuracy
            macro avg
                            0.62
                                       0.61
                                                 0.61
                                                             21
                                                 0.61
                                                             21
         weighted avg
                            0.62
                                       0.62
    print("confusion matrix for lr:", confusion_matrix(y_test, y_pred_lr))
    print("confusion matrix for dt:", confusion_matrix(y_test, y_pred_dt))
    → confusion matrix for lr: [[4 7]
          [6 4]]
         confusion matrix for dt: [[8 3]
          [5 5]]
    plt.figure(figsize=(5,5))
    plt.scatter(x=y_test,y=y_pred_lr, color="red", label="LR")
    plt.xlabel('y_test')
    plt.ylabel('y_pred')
    plt.title('lr vs dt')
    plt.legend()
    plt.show()
    # use threshold <=0.5 ==> No else if >0.5 ==> yes
    \overline{z}
                                       Ir vs dt
             1.0
```



```
plt.figure(figsize=(5,5))
plt.scatter(x=y_test,y=y_pred_dt, color="blue", label="DT")
plt.xlabel('y_test')
plt.ylabel('y_pred')
plt.title('lr vs dt')
plt.legend()
plt.show()
# use threshold <=0.5 ==> No else if >0.5 ==> yes
```



from sklearn.model\_selection import GridSearchCV, RandomizedSearchCV

```
grid_params = {"max_iter": [1000, 2000, 3000, 4000, 5000, 6000, 7000, 8000, 9000, 10000]}
grid_search = GridSearchCV(lr, grid_params, cv=5)
grid_search.fit(x_train, y_train)
print("Best parameters:", grid_search.best_params_)
print("Best score:", grid_search.best_score_)
    Best parameters: {'max_iter': 1000}
    Best score: 0.5566176470588236
random_search = RandomizedSearchCV(dt, random_params, cv=5)
random_search.fit(x_train, y_train)
print("Best parameters:", random_search.best_params_)
print("Best score:", random_search.best_score_)
    Best parameters: {'min_samples_split': 2, 'max_depth': 10}
    Best score: 0.5801470588235295
pip install gradio
→ Collecting gradio
      Downloading gradio-5.12.0-py3-none-any.whl.metadata (16 kB)
    Collecting aiofiles<24.0,>=22.0 (from gradio)
      Downloading aiofiles-23.2.1-py3-none-any.whl.metadata (9.7 kB)
    Requirement already satisfied: anyio<5.0,>=3.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (3.7.1)
    Collecting fastapi<1.0,>=0.115.2 (from gradio)
       Downloading fastapi-0.115.6-py3-none-any.whl.metadata (27 kB)
    Collecting ffmpv (from gradio)
      Downloading ffmpy-0.5.0-py3-none-any.whl.metadata (3.0 kB)
    Collecting gradio-client==1.5.4 (from gradio)
       Downloading gradio_client-1.5.4-py3-none-any.whl.metadata (7.1 kB)
    Requirement already satisfied: httpx>=0.24.1 in /usr/local/lib/python3.10/dist-packages (from gradio) (0.28.1)
    Requirement already satisfied: huggingface-hub>=0.25.1 in /usr/local/lib/python3.10/dist-packages (from gradio) (0.27.0)
    Requirement already satisfied: jinja2<4.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (3.1.4)
    Collecting markupsafe~=2.0 (from gradio)
      Downloading \ MarkupSafe-2.1.5-cp310-cp310-manylinux \\ 2\_17\_x86\_64.manylinux \\ 2014\_x86\_64.whl.metadata \ (3.0 \ kB)
    Requirement already satisfied: numpy<3.0,>=1.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (1.26.4)
    Requirement already satisfied: orjson~=3.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (3.10.12)
    Requirement already satisfied: packaging in /usr/local/lib/python3.10/dist-packages (from gradio) (24.2)
    Requirement already satisfied: pandas<3.0,>=1.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (2.2.2)
    Requirement already satisfied: pillow<12.0,>=8.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (11.0.0)
    Requirement already satisfied: pydantic>=2.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (2.10.3)
    Collecting pydub (from gradio)
      Downloading pydub-0.25.1-py2.py3-none-any.whl.metadata (1.4 kB)
    Collecting python-multipart>=0.0.18 (from gradio)
      Downloading python_multipart-0.0.20-py3-none-any.whl.metadata (1.8 kB)
    Requirement already satisfied: pyyaml<7.0,>=5.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (6.0.2)
    Collecting ruff>=0.2.2 (from gradio)
```

```
Downloading ruff-0.9.1-py3-none-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (25 kB)
Collecting safehttpx<0.2.0,>=0.1.6 (from gradio)
 Downloading safehttpx-0.1.6-py3-none-any.whl.metadata (4.2 kB)
Collecting semantic-version~=2.0 (from gradio)
 Downloading semantic_version-2.10.0-py2.py3-none-any.whl.metadata (9.7 kB)
Collecting starlette<1.0,>=0.40.0 (from gradio)
 Downloading starlette-0.45.2-py3-none-any.whl.metadata (6.3 kB)
Collecting tomlkit<0.14.0,>=0.12.0 (from gradio)
 Downloading tomlkit-0.13.2-py3-none-any.whl.metadata (2.7 kB)
Requirement already satisfied: typer<1.0,>=0.12 in /usr/local/lib/python3.10/dist-packages (from gradio) (0.15.1)
Requirement already satisfied: typing-extensions~=4.0 in /usr/local/lib/python3.10/dist-packages (from gradio) (4.12.2)
Collecting uvicorn>=0.14.0 (from gradio)
 Downloading uvicorn-0.34.0-py3-none-any.whl.metadata (6.5 kB)
Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages (from gradio-client==1.5.4->gradio) (2024.10.0)
Requirement already satisfied: websockets<15.0,>=10.0 in /usr/local/lib/python3.10/dist-packages (from gradio-client==1.5.4->gradio)
Requirement already satisfied: idna>=2.8 in /usr/local/lib/python3.10/dist-packages (from anyio<5.0,>=3.0->gradio) (3.10)
Requirement already satisfied: sniffio>=1.1 in /usr/local/lib/python3.10/dist-packages (from anyio<5.0,>=3.0->gradio) (1.3.1)
Requirement already satisfied: exceptiongroup in /usr/local/lib/python3.10/dist-packages (from anyio<5.0,>=3.0->gradio) (1.2.2)
Collecting starlette<1.0,>=0.40.0 (from gradio)
 Downloading starlette-0.41.3-py3-none-any.whl.metadata (6.0 kB)
Requirement already satisfied: certifi in /usr/local/lib/python3.10/dist-packages (from httpx>=0.24.1->gradio) (2024.12.14)
Requirement already satisfied: httpcore==1.* in /usr/local/lib/python3.10/dist-packages (from httpx>=0.24.1->gradio) (1.0.7)
Requirement already satisfied: h11<0.15,>=0.13 in /usr/local/lib/python3.10/dist-packages (from httpcore==1.*->httpx>=0.24.1->gradio)
Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-packages (from huggingface-hub>=0.25.1->gradio) (3.16.1)
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from huggingface-hub>=0.25.1->gradio) (2.32.3)
Requirement already satisfied: tqdm>=4.42.1 in /usr/local/lib/python3.10/dist-packages (from huggingface-hub>=0.25.1->gradio) (4.67.1)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.10/dist-packages (from pandas<3.0,>=1.0->gradio) (2.8
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas<3.0,>=1.0->gradio) (2024.2)
Requirement already satisfied: tzdatas=2022.7 in /usr/local/lih/nython3.10/dist-nackages (from nandas<3.0.5=1.0-5gradio) (2024.2)
```

import gradio as gr