

In [2]: `!pip install numpy pandas matplotlib seaborn scikit-p learn`

```
Requirement already satisfied: numpy in c:\users\sudha\anaconda3\lib\site-packages (1.26.4)
Requirement already satisfied: pandas in c:\users\sudha\anaconda3\lib\site-packages (2.2.2)
Requirement already satisfied: matplotlib in c:\users\sudha\anaconda3\lib\site-packages (3.9.2)
Requirement already satisfied: seaborn in c:\users\sudha\anaconda3\lib\site-packages (0.13.2)
Requirement already satisfied: scikit-learn in c:\users\sudha\anaconda3\lib\site-packages (1.5.1)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\sudha\anaconda3\lib\site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\sudha\anaconda3\lib\site-packages (from pandas) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in c:\users\sudha\anaconda3\lib\site-packages (from pandas) (2023.3)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\sudha\anaconda3\lib\site-packages (from matplotlib) (1.2.0)
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Requirement already satisfied: fonttools>=4.22.0 in c:\users\sudha\anaconda3\lib\site-packages (from matplotlib) (4.51.0)
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Requirement already satisfied: pillow>=8 in c:\users\sudha\anaconda3\lib\site-packages (from matplotlib) (10.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\sudha\anaconda3\lib\site-packages (from matplotlib) (3.1.2)
Requirement already satisfied: scipy>=1.6.0 in c:\users\sudha\anaconda3\lib\site-packages (from scikit-learn) (1.13.1)
Requirement already satisfied: joblib>=1.2.0 in c:\users\sudha\anaconda3\lib\site-packages (from scikit-learn) (1.4.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in c:\users\sudha\anaconda3\lib\site-packages (from scikit-learn) (3.5.0)
Requirement already satisfied: six>=1.5 in c:\users\sudha\anaconda3\lib\site-packages (from python-dateutil>=2.8.2->pandas) (1.16.0)
```

In [4]: `import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error, r2_score`

In [6]: `np.random.seed(42)
data = {
 'Latency (ms)': np.random.randint(20, 300, 1000),
 'Packet Loss (%)': np.random.uniform(0, 5, 1000),
 'Jitter (ms)': np.random.randint(5, 100, 1000),
 'Bandwidth (Mbps)': np.random.uniform(5, 100, 1000),
}
data['QoE Score'] = 5 - (data['Latency (ms)'] / 80 + data['Packet Loss (%)'] / 2 + dat`

```
data['QoE Score'] = np.clip(data['QoE Score'], 1, 5)
df = pd.DataFrame(data)
```

```
In [8]: print("First 5 rows of dataset:")
print(df.head())
print("\nSummary Statistics:")
print(df.describe())
```

First 5 rows of dataset:

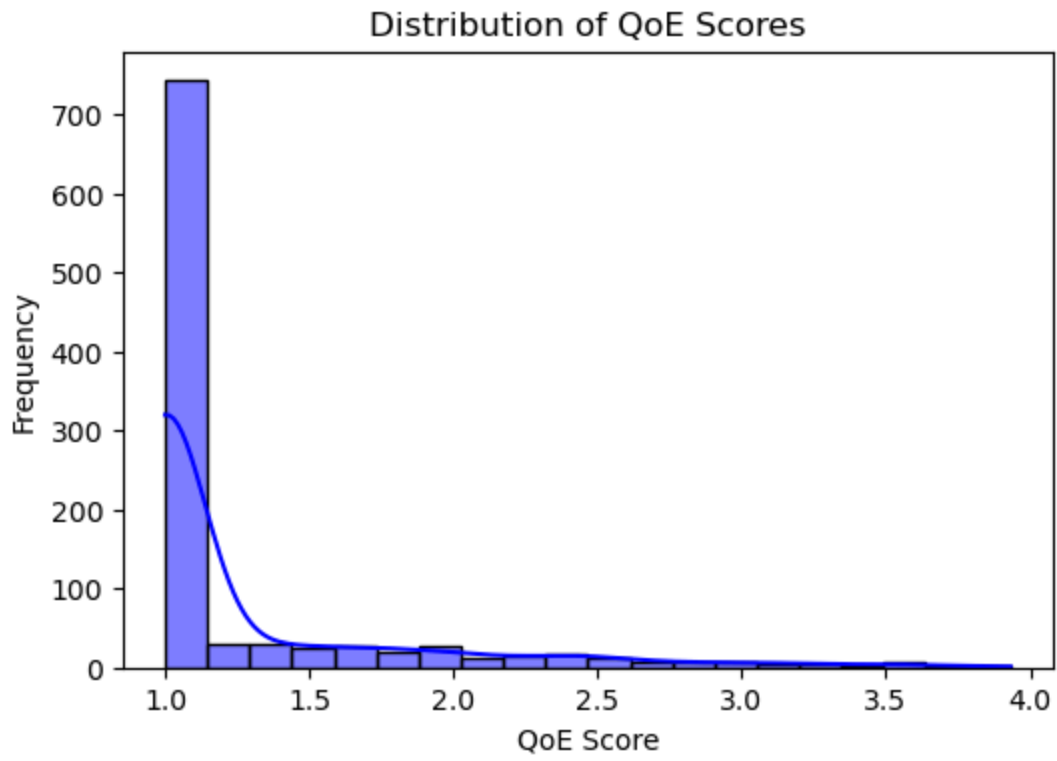
	Latency (ms)	Packet Loss (%)	Jitter (ms)	Bandwidth (Mbps)	QoE Score
0	122	4.472761	42	54.416140	1.000000
1	290	3.999276	50	18.573216	1.000000
2	126	2.126068	8	78.657884	2.095300
3	91	0.112347	64	30.783891	1.672993
4	208	1.343387	61	52.186065	1.000000

Summary Statistics:

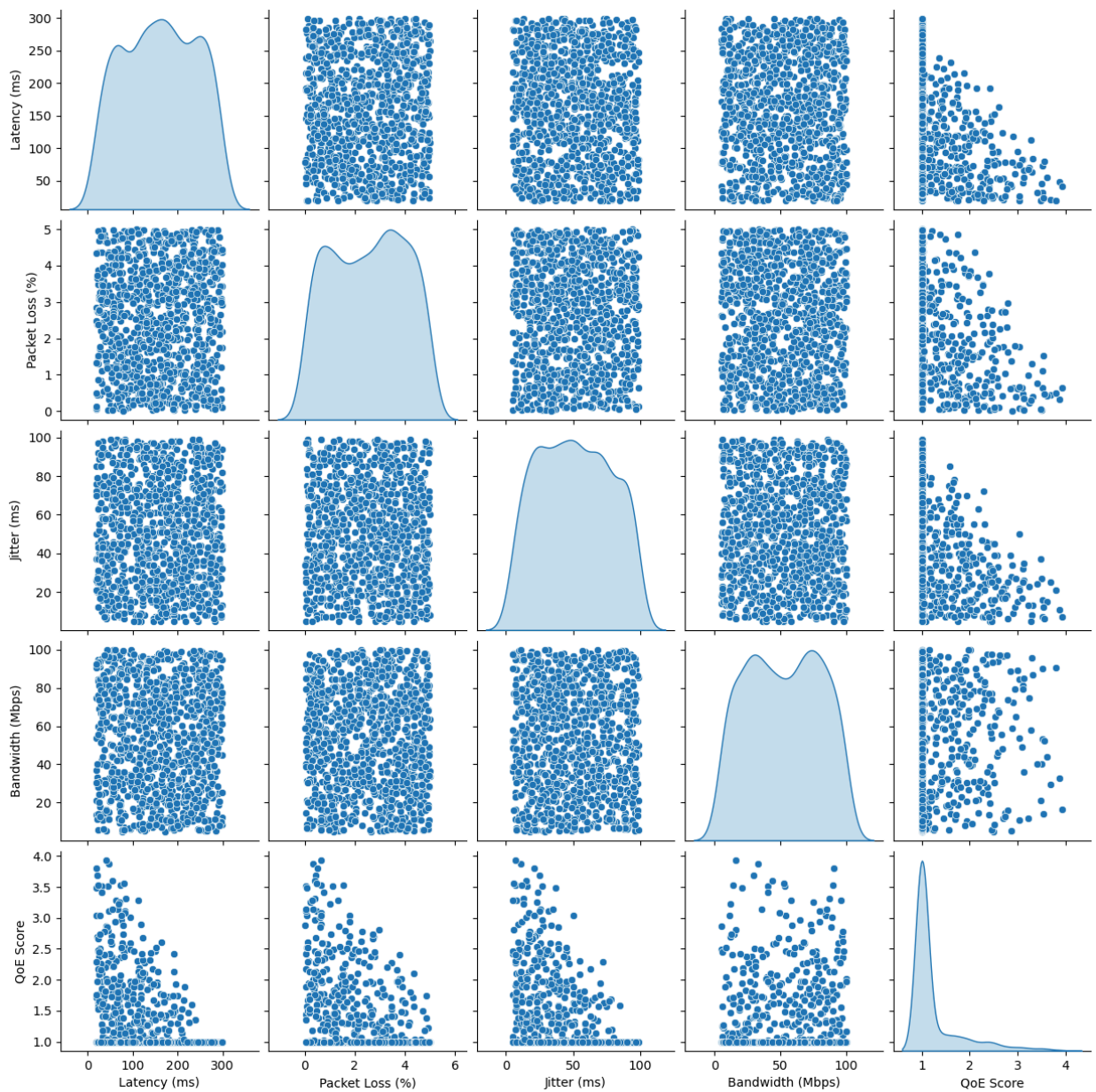
	Latency (ms)	Packet Loss (%)	Jitter (ms)	Bandwidth (Mbps)	\
count	1000.000000	1000.000000	1000.000000	1000.000000	
mean	159.911000	2.529291	51.287000	52.266419	
std	79.781452	1.466028	26.553036	27.456459	
min	20.000000	0.016091	5.000000	5.001105	
25%	91.750000	1.201573	28.000000	28.640983	
50%	162.500000	2.594944	51.000000	52.522429	
75%	231.250000	3.793849	73.000000	76.081555	
max	299.000000	4.997069	99.000000	99.894798	

	QoE Score
count	1000.000000
mean	1.258111
std	0.544678
min	1.000000
25%	1.000000
50%	1.000000
75%	1.179662
max	3.931992

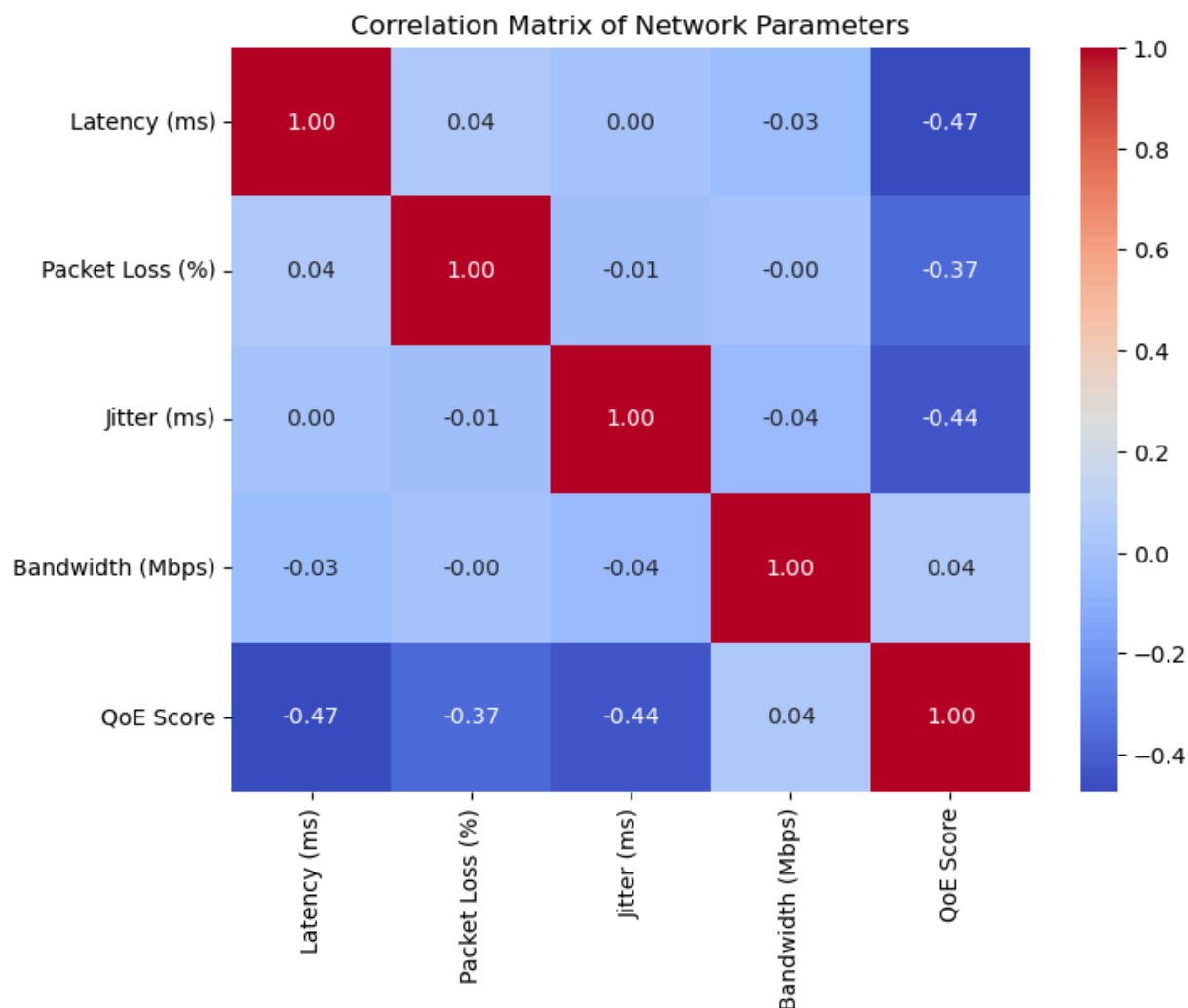
```
In [10]: plt.figure(figsize=(6,4))
sns.histplot(df['QoE Score'], bins=20, kde=True, color='blue')
plt.title("Distribution of QoE Scores")
plt.xlabel("QoE Score")
plt.ylabel("Frequency")
plt.show()
```



```
In [12]: sns.pairplot(df, diag_kind='kde')  
plt.show()
```



```
In [14]: plt.figure(figsize=(8,6))
sns.heatmap(df.corr(), annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Matrix of Network Parameters")
plt.show()
```



```
In [16]: X = df[['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)', 'Bandwidth (Mbps)']]
y = df['QoE Score']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=
```

```
In [18]: lr_model = LinearRegression()
lr_model.fit(X_train, y_train)
y_pred_lr = lr_model.predict(X_test)
mse_lr = mean_squared_error(y_test, y_pred_lr)
r2_lr = r2_score(y_test, y_pred_lr)
print(f"Linear Regression - MSE: {mse_lr:.2f}, R2 Score: {r2_lr:.2f}")
```

Linear Regression - MSE: 0.12, R2 Score: 0.55

```
In [20]: rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)
y_pred_rf = rf_model.predict(X_test)
mse_rf = mean_squared_error(y_test, y_pred_rf)
r2_rf = r2_score(y_test, y_pred_rf)
print(f"Random Forest - MSE: {mse_rf:.2f}, R2 Score: {r2_rf:.2f}")
```

Random Forest - MSE: 0.01, R2 Score: 0.96

```
In [22]: models = ["Linear Regression", "Random Forest"]
mse_values = [mse_lr, mse_rf]
r2_values = [r2_lr, r2_rf]
```

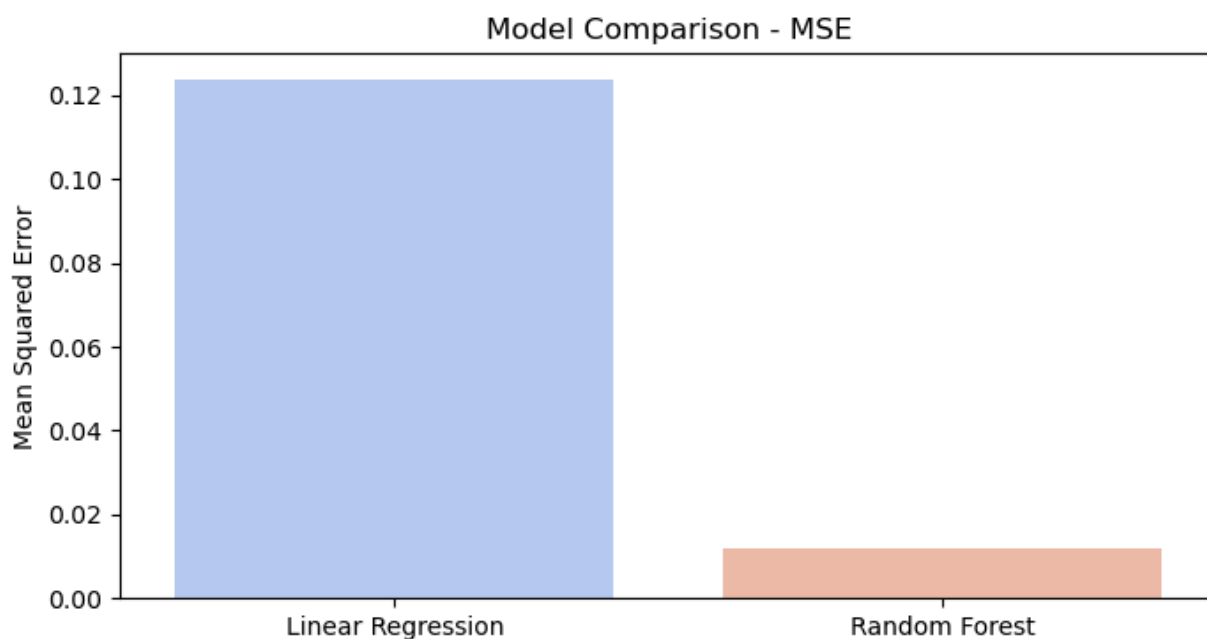
```
plt.figure(figsize=(8,4))
sns.barplot(x=models, y=mse_values, palette='coolwarm')
plt.title("Model Comparison - MSE")
plt.ylabel("Mean Squared Error")
plt.show()

plt.figure(figsize=(8,4))
sns.barplot(x=models, y=r2_values, palette='coolwarm')
plt.title("Model Comparison - R2 Score")
plt.ylabel("R2 Score")
plt.show()
```

C:\Users\sudha\AppData\Local\Temp\ipykernel\_1164\1199408385.py:6: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

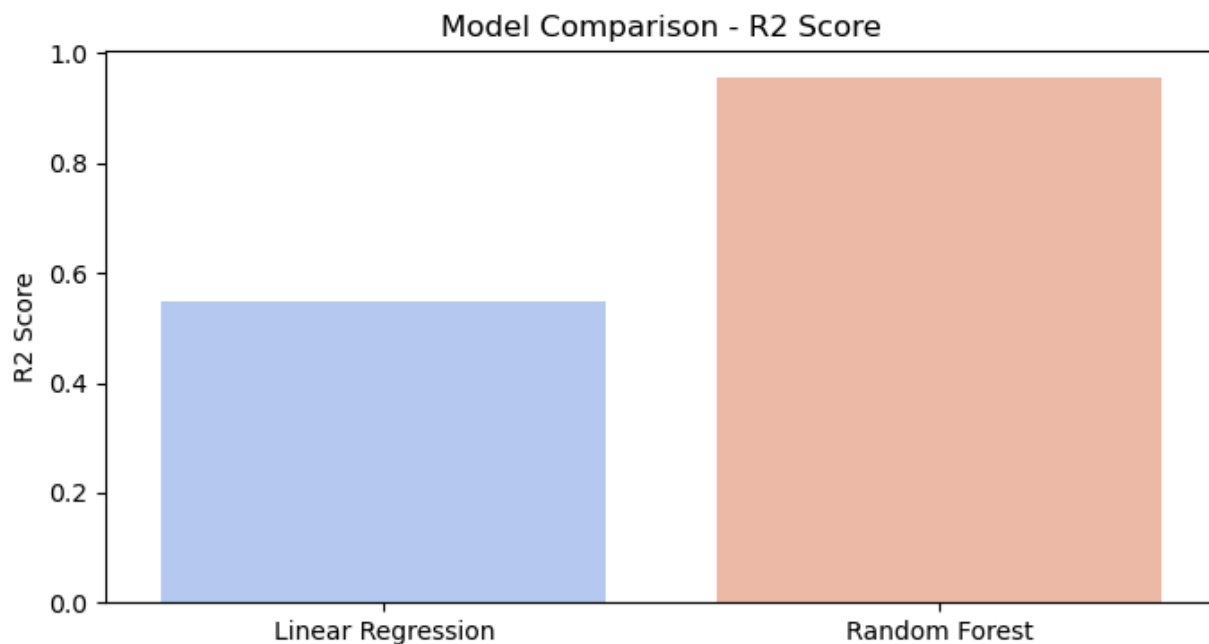
```
sns.barplot(x=models, y=mse_values, palette='coolwarm')
```



C:\Users\sudha\AppData\Local\Temp\ipykernel\_1164\1199408385.py:12: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=models, y=r2_values, palette='coolwarm')
```

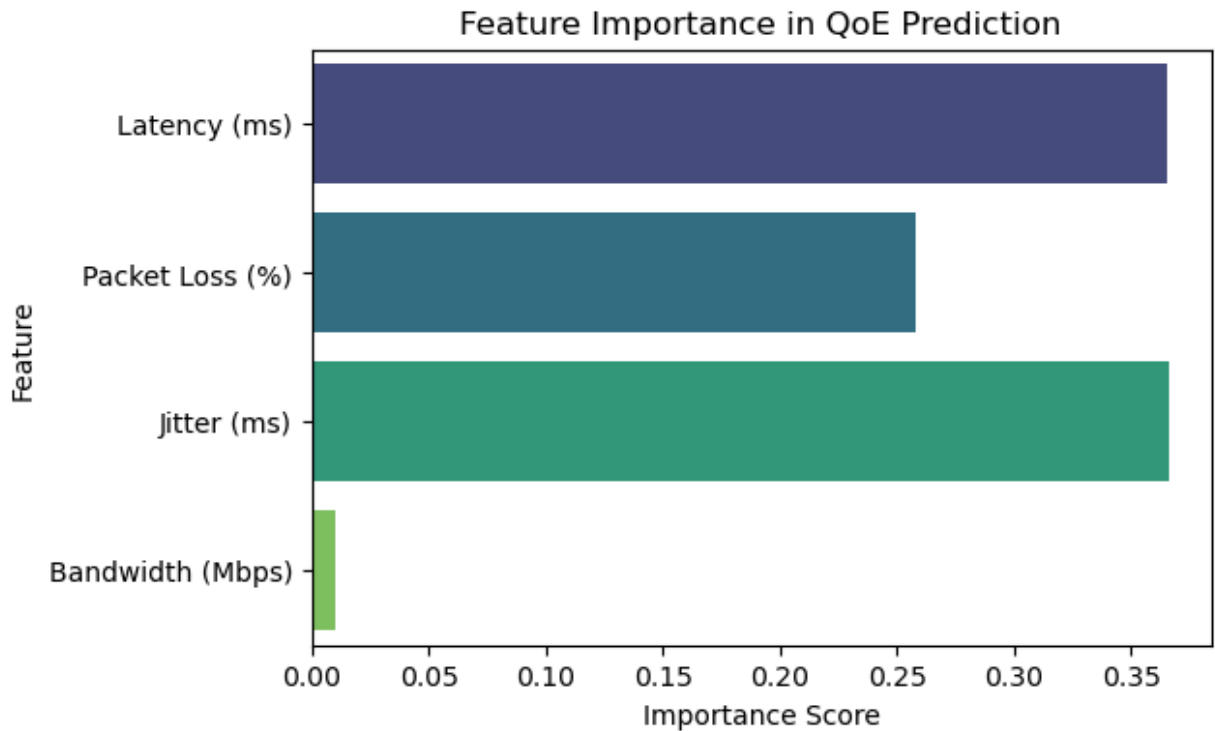


```
In [24]: importance = rf_model.feature_importances_
features = X.columns
plt.figure(figsize=(6,4))
sns.barplot(x=importance, y=features, palette='viridis')
plt.title("Feature Importance in QoE Prediction")
plt.xlabel("Importance Score")
plt.ylabel("Feature")
plt.show()
```

C:\Users\sudha\AppData\Local\Temp\ipykernel\_1164\3597949485.py:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=importance, y=features, palette='viridis')
```



```
In [26]: optimized_conditions = pd.DataFrame({
    'Latency (ms)': np.linspace(20, 100, 10),
    'Packet Loss (%)': np.linspace(0, 2, 10),
    'Jitter (ms)': np.linspace(5, 30, 10),
    'Bandwidth (Mbps)': np.linspace(10, 100, 10)
})
optimized_conditions['Predicted QoE'] = rf_model.predict(optimized_conditions)
best_conditions = optimized_conditions.loc[optimized_conditions['Predicted QoE'].idxmax()]
print("Optimal Network Conditions for Best QoE:")
print(best_conditions)
```

```
Optimal Network Conditions for Best QoE:
Latency (ms)      28.888889
Packet Loss (%)   0.222222
Jitter (ms)       7.777778
Bandwidth (Mbps)  20.000000
Predicted QoE     3.776261
Name: 1, dtype: float64
```

```
In [28]: def predict_qoe(latency, packet_loss, jitter, bandwidth):
    input_data = pd.DataFrame([[latency, packet_loss, jitter, bandwidth]],
                               columns=['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)', 'Bandwidth (Mbps)'])
    qoe_score = rf_model.predict(input_data)[0]
    return round(qoe_score, 2)
```

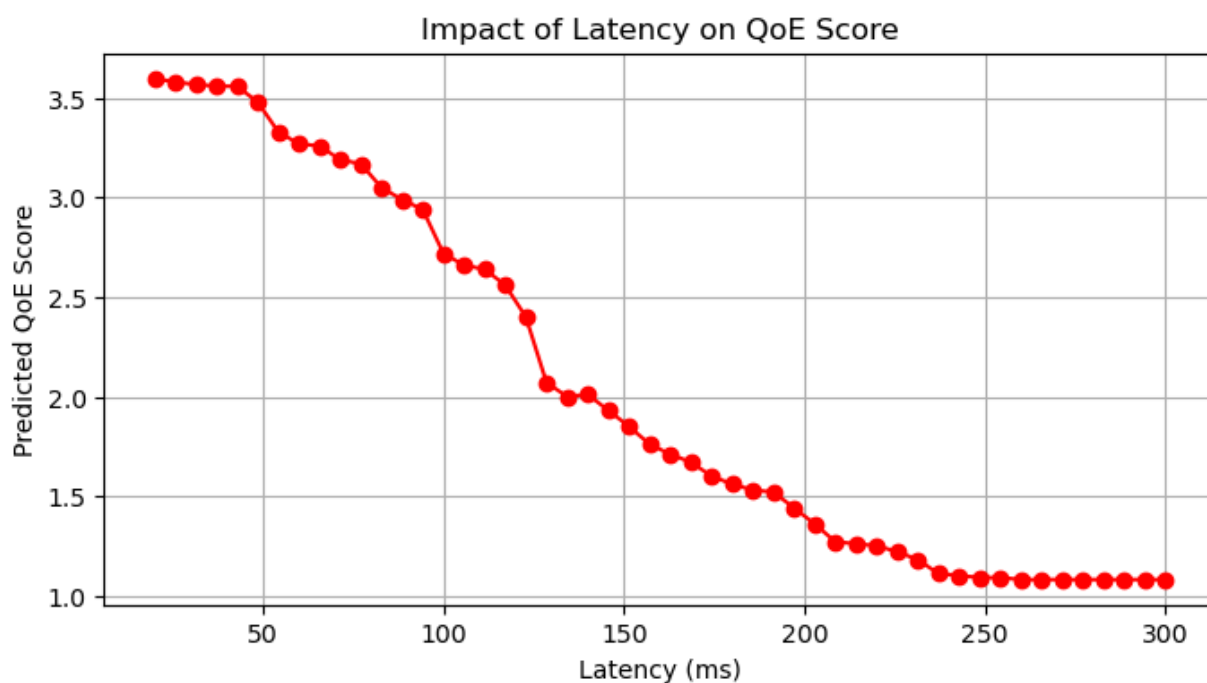
```
latency = 50
packet_loss = 1
jitter = 10
bandwidth = 25
predicted_qoe = predict_qoe(latency, packet_loss, jitter, bandwidth)
print(f"Predicted QoE Score: {predicted_qoe}/5")
```

```
Predicted QoE Score: 3.47/5
```

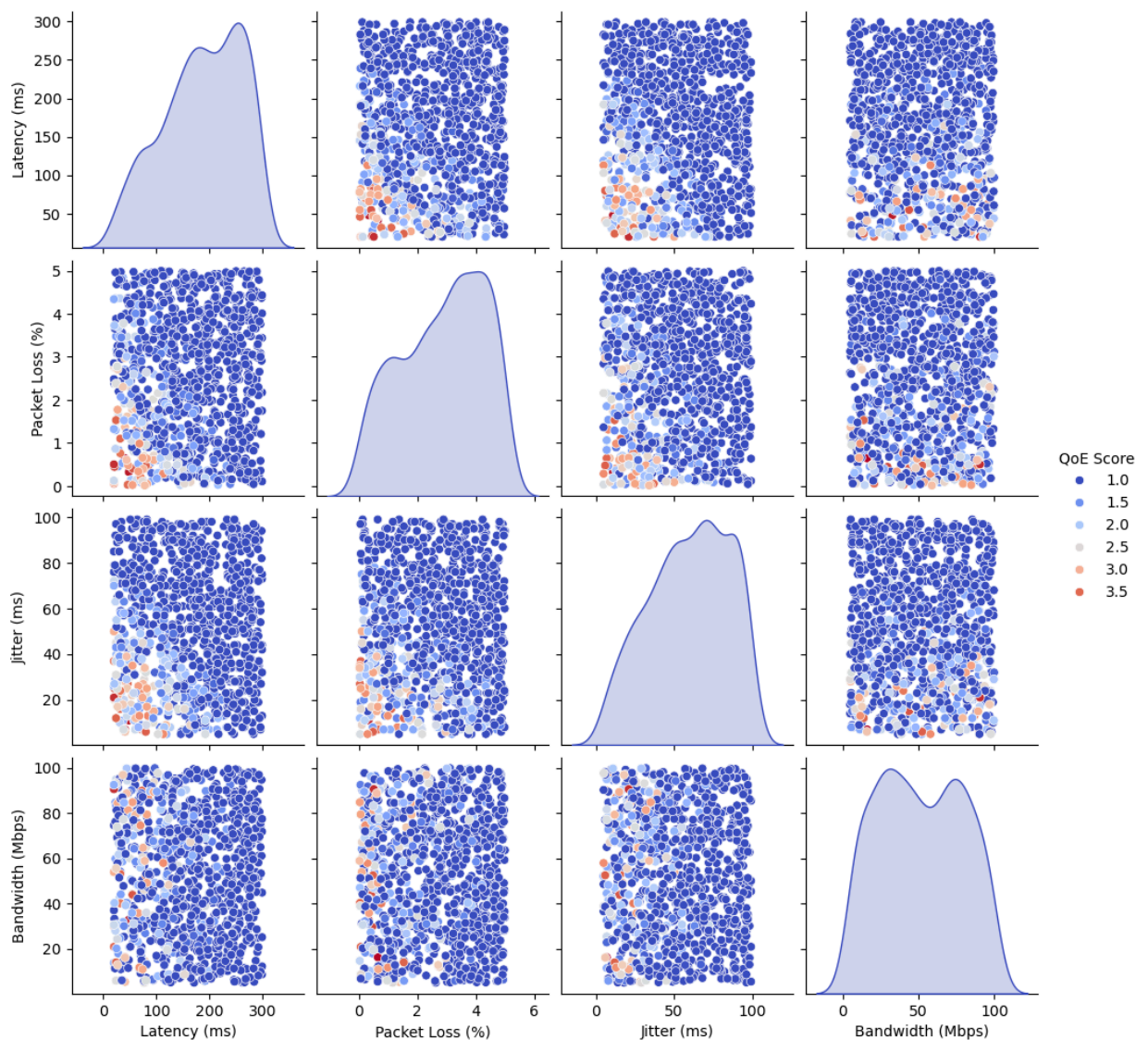
```
In [30]: latencies = np.linspace(20, 300, 50)
predicted_qoe_values = [predict_qoe(lat, 1, 10, 50) for lat in latencies]
```



```
plt.figure(figsize=(8,4))
plt.plot(latencies, predicted_qoe_values, marker='o', linestyle='-', color='red')
plt.title("Impact of Latency on QoE Score")
plt.xlabel("Latency (ms)")
plt.ylabel("Predicted QoE Score")
plt.grid()
plt.show()
```



```
In [32]: sns.pairplot(df, hue="QoE Score", palette="coolwarm")
plt.show()
```

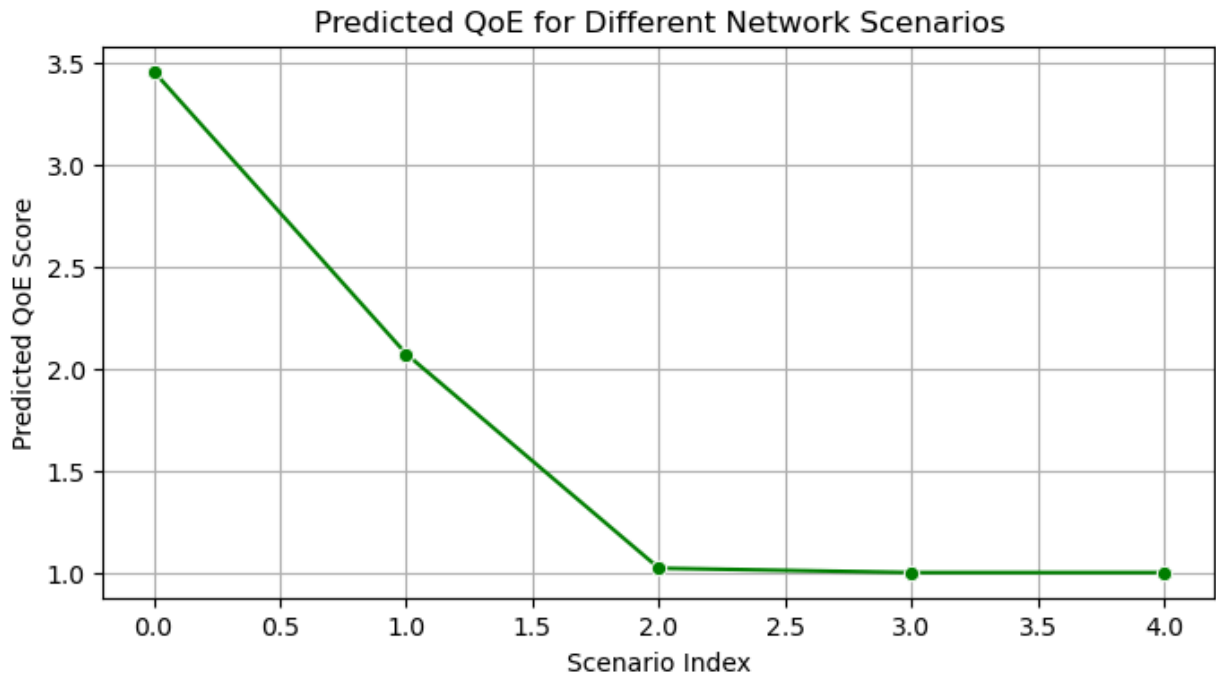


```
In [34]: scenarios = pd.DataFrame({
    'Latency (ms)': [50, 100, 150, 200, 250],
    'Packet Loss (%)': [1, 2, 3, 4, 5],
    'Jitter (ms)': [10, 20, 30, 40, 50],
    'Bandwidth (Mbps)': [10, 20, 30, 40, 50]
})
scenarios['Predicted QoE'] = rf_model.predict(scenarios)
print("Predicted QoE Scores for Various Scenarios:")
print(scenarios)

plt.figure(figsize=(8,4))
sns.lineplot(x=scenarios.index, y=scenarios['Predicted QoE'], marker='o', color='green')
plt.title("Predicted QoE for Different Network Scenarios")
plt.xlabel("Scenario Index")
plt.ylabel("Predicted QoE Score")
plt.grid()
plt.show()
```

Predicted QoE Scores for Various Scenarios:

	Latency (ms)	Packet Loss (%)	Jitter (ms)	Bandwidth (Mbps)	Predicted QoE
0	50	1	10	10	3.457042
1	100	2	20	20	2.071310
2	150	3	30	30	1.021342
3	200	4	40	40	1.000000
4	250	5	50	50	1.000000



```
In [2]: !pip install tensorflow
```

Collecting tensorflow  
 Downloading tensorflow-2.19.0-cp312-cp312-win\_amd64.whl.metadata (4.1 kB)  
Collecting absl-py>=1.0.0 (from tensorflow)  
 Downloading absl\_py-2.2.0-py3-none-any.whl.metadata (2.4 kB)  
Collecting astunparse>=1.6.0 (from tensorflow)  
 Downloading astunparse-1.6.3-py2.py3-none-any.whl.metadata (4.4 kB)  
Collecting flatbuffers>=24.3.25 (from tensorflow)  
 Downloading flatbuffers-25.2.10-py2.py3-none-any.whl.metadata (875 bytes)  
Collecting gast!=0.5.0,!0.5.1,!0.5.2,>=0.2.1 (from tensorflow)  
 Downloading gast-0.6.0-py3-none-any.whl.metadata (1.3 kB)  
Collecting google-pasta>=0.1.1 (from tensorflow)  
 Downloading google\_pasta-0.2.0-py3-none-any.whl.metadata (814 bytes)  
Collecting libclang>=13.0.0 (from tensorflow)  
 Downloading libclang-18.1.1-py2.py3-none-win\_amd64.whl.metadata (5.3 kB)  
Collecting opt-einsum>=2.3.2 (from tensorflow)  
 Downloading opt\_einsum-3.4.0-py3-none-any.whl.metadata (6.3 kB)  
Requirement already satisfied: packaging in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (24.1)  
Requirement already satisfied: protobuf!=4.21.0,!4.21.1,!4.21.2,!4.21.3,!4.21.4,!4.21.5,<6.0.0dev,>=3.20.3 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (4.25.3)  
Requirement already satisfied: requests<3,>=2.21.0 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (2.32.3)  
Requirement already satisfied: setuptools in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (75.1.0)  
Requirement already satisfied: six>=1.12.0 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (1.16.0)  
Collecting termcolor>=1.1.0 (from tensorflow)  
 Downloading termcolor-2.5.0-py3-none-any.whl.metadata (6.1 kB)  
Requirement already satisfied: typing-extensions>=3.6.6 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (4.11.0)  
Requirement already satisfied: wrapt>=1.11.0 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (1.14.1)  
Collecting grpcio<2.0,>=1.24.3 (from tensorflow)  
 Downloading grpcio-1.71.0-cp312-cp312-win\_amd64.whl.metadata (4.0 kB)  
Collecting tensorboard~2.19.0 (from tensorflow)  
 Downloading tensorboard-2.19.0-py3-none-any.whl.metadata (1.8 kB)  
Collecting keras>=3.5.0 (from tensorflow)  
 Downloading keras-3.9.0-py3-none-any.whl.metadata (6.1 kB)  
Requirement already satisfied: numpy<2.2.0,>=1.26.0 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (1.26.4)  
Requirement already satisfied: h5py>=3.11.0 in c:\users\sudha\anaconda3\lib\site-packages (from tensorflow) (3.11.0)  
Collecting ml-dtypes<1.0.0,>=0.5.1 (from tensorflow)  
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Requirement already satisfied: wheel<1.0,>=0.23.0 in c:\users\sudha\anaconda3\lib\site-packages (from astunparse>=1.6.0->tensorflow) (0.44.0)  
Requirement already satisfied: rich in c:\users\sudha\anaconda3\lib\site-packages (from keras>=3.5.0->tensorflow) (13.7.1)  
Collecting namex (from keras>=3.5.0->tensorflow)  
 Downloading namex-0.0.8-py3-none-any.whl.metadata (246 bytes)  
Collecting optree (from keras>=3.5.0->tensorflow)  
 Downloading optree-0.14.1-cp312-cp312-win\_amd64.whl.metadata (50 kB)  
Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\sudha\anaconda3\lib\site-packages (from requests<3,>=2.21.0->tensorflow) (3.3.2)  
Requirement already satisfied: idna<4,>=2.5 in c:\users\sudha\anaconda3\lib\site-packages (from requests<3,>=2.21.0->tensorflow) (3.7)  
Requirement already satisfied: urllib3<3,>=1.21.1 in c:\users\sudha\anaconda3\lib\site-packages (from requests<3,>=2.21.0->tensorflow) (2.2.3)  
Requirement already satisfied: certifi>=2017.4.17 in c:\users\sudha\anaconda3\lib\site-packages (from requests<3,>=2.21.0->tensorflow) (2025.1.31)

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Requirement already satisfied: markdown>=2.6.8 in c:\users\sudha\anaconda3\lib\site-p
ackages (from tensorboard~=2.19.0->tensorflow) (3.4.1)
Collecting tensorboard-data-server<0.8.0,>=0.7.0 (from tensorboard~=2.19.0->tensorflo
w)
  Downloading tensorboard_data_server-0.7.2-py3-none-any.whl.metadata (1.1 kB)
Requirement already satisfied: werkzeug>=1.0.1 in c:\users\sudha\anaconda3\lib\site-p
ackages (from tensorboard~=2.19.0->tensorflow) (3.0.3)
Requirement already satisfied: MarkupSafe>=2.1.1 in c:\users\sudha\anaconda3\lib\site
-packages (from werkzeug>=1.0.1->tensorboard~=2.19.0->tensorflow) (2.1.3)
Requirement already satisfied: markdown-it-py>=2.2.0 in c:\users\sudha\anaconda3\lib
\site-packages (from rich->keras>=3.5.0->tensorflow) (2.2.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in c:\users\sudha\anaconda3\li
b\site-packages (from rich->keras>=3.5.0->tensorflow) (2.15.1)
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es (from markdown-it-py>=2.2.0->rich->keras>=3.5.0->tensorflow) (0.1.0)
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----- 0.8/376.0 MB 5.6 MB/s eta 0:01:08
----- 1.3/376.0 MB 3.9 MB/s eta 0:01:36
----- 1.8/376.0 MB 3.6 MB/s eta 0:01:45
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----- 65.3/376.0 MB 11.4 MB/s eta 0:00:28
----- 68.7/376.0 MB 11.5 MB/s eta 0:00:27
----- 71.0/376.0 MB 11.6 MB/s eta 0:00:27
----- 74.2/376.0 MB 11.7 MB/s eta 0:00:26
----- 76.0/376.0 MB 11.6 MB/s eta 0:00:26
----- 77.6/376.0 MB 11.5 MB/s eta 0:00:26
----- 79.7/376.0 MB 11.4 MB/s eta 0:00:26
----- 81.0/376.0 MB 11.3 MB/s eta 0:00:27
----- 81.8/376.0 MB 11.1 MB/s eta 0:00:27
----- 82.3/376.0 MB 10.9 MB/s eta 0:00:27
----- 82.8/376.0 MB 10.7 MB/s eta 0:00:28
----- 84.1/376.0 MB 10.5 MB/s eta 0:00:28
----- 84.7/376.0 MB 10.4 MB/s eta 0:00:28
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----- 91.0/376.0 MB 9.9 MB/s eta 0:00:29
----- 92.3/376.0 MB 9.8 MB/s eta 0:00:30
----- 93.1/376.0 MB 9.7 MB/s eta 0:00:30
----- 93.6/376.0 MB 9.5 MB/s eta 0:00:30
----- 93.8/376.0 MB 9.4 MB/s eta 0:00:31
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Downloading opt_einsum-3.4.0-py3-none-any.whl (71 kB)
Downloading tensorboard-2.19.0-py3-none-any.whl (5.5 MB)
----- 0.0/5.5 MB ? eta -:--:--
----- 1.6/5.5 MB 8.4 MB/s eta 0:00:01
----- 3.4/5.5 MB 7.7 MB/s eta 0:00:01
----- 5.2/5.5 MB 8.0 MB/s eta 0:00:01
----- 5.5/5.5 MB 7.8 MB/s eta 0:00:00
Downloading termcolor-2.5.0-py3-none-any.whl (7.8 kB)
Downloading tensorboard_data_server-0.7.2-py3-none-any.whl (2.4 kB)
Downloading namex-0.0.8-py3-none-any.whl (5.8 kB)
Downloading optree-0.14.1-cp312-cp312-win_amd64.whl (306 kB)
Installing collected packages: namex, libclang, flatbuffers, termcolor, tensorboard-ata-server, optree, opt-einsum, ml-dtypes, grpcio, google-pasta, gast, astunparse, absl-py, tensorboard, keras, tensorflow

```

Successfully installed absl-py-2.2.0 astunparse-1.6.3 flatbuffers-25.2.10 gast-0.6.0 google-pasta-0.2.0 grpcio-1.71.0 keras-3.9.0 libclang-18.1.1 ml-dtypes-0.5.1 namex-0.0.8 opt-einsum-3.4.0 optree-0.14.1 tensorboard-2.19.0 tensorboard-data-server-0.7.2 tensorflow-2.19.0 termcolor-2.5.0

```
In [6]: import numpy as np
import pandas as pd

# Simulate Network Data
np.random.seed(42)
data = {
    'Latency (ms)': np.random.randint(20, 300, 1000),
    'Packet Loss (%)': np.random.uniform(0, 5, 1000),
    'Jitter (ms)': np.random.randint(5, 100, 1000),
    'Bandwidth (Mbps)': np.random.uniform(5, 100, 1000),
}
data['QoE Score'] = 5 - (data['Latency (ms)'] / 80 + data['Packet Loss (%)'] / 2 + data['Jitter (ms)'] / 100)
data['QoE Score'] = np.clip(data['QoE Score'], 1, 5)

# Create DataFrame
df = pd.DataFrame(data)
```

```
In [8]: import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Input





















# Ensure X_train is defined before running this step
if 'X_train' not in globals():
    from sklearn.model_selection import train_test_split
    X = df[['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)', 'Bandwidth (Mbps)']]
    y = df['QoE Score']
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)





















# Define a deep learning model
deep_model = Sequential([
    Input(shape=(4,)), # Correct way to specify input shape
    Dense(64, activation='relu'),
    Dense(32, activation='relu'),
    Dense(16, activation='relu'),
    Dense(1) # Output Layer (QoE Score)
])














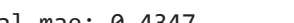

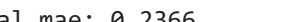




# Compile the model
deep_model.compile(optimizer='adam', loss='mse', metrics=['mae'])













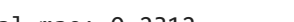
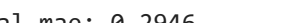
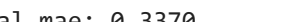
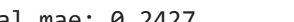
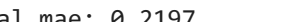



# Train the model
deep_model.fit(X_train, y_train, epochs=100, batch_size=16, verbose=1, validation_data=(X_test, y_test))

# Evaluate the model
dl_loss, dl_mae = deep_model.evaluate(X_test, y_test)
print(f"Deep Learning Model - Loss: {dl_loss:.2f}, MAE: {dl_mae:.2f}")
```














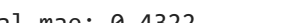

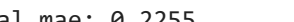
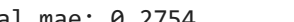

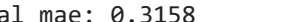

Epoch 1/100  
50/50  2s 8ms/step - loss: 304.2001 - mae: 12.3210 - val\_loss: 4.2631 - val\_mae: 1.6657  
Epoch 2/100  
50/50  0s 4ms/step - loss: 2.5193 - mae: 1.2723 - val\_loss: 1.0050 - val\_mae: 0.7468  
Epoch 3/100  
50/50  0s 4ms/step - loss: 0.8398 - mae: 0.6800 - val\_loss: 0.6221 - val\_mae: 0.5805  
Epoch 4/100  
50/50  0s 4ms/step - loss: 0.5708 - mae: 0.5357 - val\_loss: 0.5037 - val\_mae: 0.4990  
Epoch 5/100  
50/50  0s 4ms/step - loss: 0.4505 - mae: 0.4451 - val\_loss: 0.4726 - val\_mae: 0.4951  
Epoch 6/100  
50/50  0s 4ms/step - loss: 0.4911 - mae: 0.4673 - val\_loss: 0.4511 - val\_mae: 0.4767  
Epoch 7/100  
50/50  0s 4ms/step - loss: 0.5117 - mae: 0.4697 - val\_loss: 0.4381 - val\_mae: 0.4597  
Epoch 8/100  
50/50  0s 4ms/step - loss: 0.5193 - mae: 0.4671 - val\_loss: 0.4373 - val\_mae: 0.4378  
Epoch 9/100  
50/50  0s 4ms/step - loss: 0.4233 - mae: 0.4174 - val\_loss: 0.4163 - val\_mae: 0.4506  
Epoch 10/100  
50/50  0s 4ms/step - loss: 0.4170 - mae: 0.3947 - val\_loss: 0.4256 - val\_mae: 0.4702  
Epoch 11/100  
50/50  0s 4ms/step - loss: 0.4422 - mae: 0.4318 - val\_loss: 0.4046 - val\_mae: 0.4366  
Epoch 12/100  
50/50  0s 4ms/step - loss: 0.4606 - mae: 0.4314 - val\_loss: 0.3872 - val\_mae: 0.4328  
Epoch 13/100  
50/50  0s 4ms/step - loss: 0.4831 - mae: 0.4285 - val\_loss: 0.4001 - val\_mae: 0.4618  
Epoch 14/100  
50/50  0s 4ms/step - loss: 0.4310 - mae: 0.4026 - val\_loss: 0.3928 - val\_mae: 0.4563  
Epoch 15/100  
50/50  0s 4ms/step - loss: 0.3279 - mae: 0.3809 - val\_loss: 0.3806 - val\_mae: 0.4362  
Epoch 16/100  
50/50  0s 4ms/step - loss: 0.3976 - mae: 0.4015 - val\_loss: 0.3700 - val\_mae: 0.4225  
Epoch 17/100  
50/50  0s 4ms/step - loss: 0.4108 - mae: 0.4146 - val\_loss: 0.3885 - val\_mae: 0.4526  
Epoch 18/100  
50/50  0s 4ms/step - loss: 0.3989 - mae: 0.4020 - val\_loss: 0.3775 - val\_mae: 0.3993  
Epoch 19/100  
50/50  0s 4ms/step - loss: 0.4455 - mae: 0.4149 - val\_loss: 0.3660 - val\_mae: 0.4182  
Epoch 20/100  
50/50  0s 4ms/step - loss: 0.3940 - mae: 0.4139 - val\_loss: 0.3537 - val\_mae: 0.3917

Epoch 21/100  
50/50  0s 4ms/step - loss: 0.3614 - mae: 0.3704 - val\_loss: 0.355  
2 - val\_mae: 0.4242  
Epoch 22/100  
50/50  0s 3ms/step - loss: 0.3502 - mae: 0.3869 - val\_loss: 0.355  
3 - val\_mae: 0.3919  
Epoch 23/100  
50/50  0s 4ms/step - loss: 0.3268 - mae: 0.3583 - val\_loss: 0.331  
7 - val\_mae: 0.3935  
Epoch 24/100  
50/50  0s 4ms/step - loss: 0.3561 - mae: 0.3988 - val\_loss: 0.323  
0 - val\_mae: 0.3690  
Epoch 25/100  
50/50  0s 4ms/step - loss: 0.2943 - mae: 0.3313 - val\_loss: 0.337  
3 - val\_mae: 0.3979  
Epoch 26/100  
50/50  0s 4ms/step - loss: 0.3187 - mae: 0.3489 - val\_loss: 0.333  
8 - val\_mae: 0.4273  
Epoch 27/100  
50/50  0s 4ms/step - loss: 0.2742 - mae: 0.3414 - val\_loss: 0.308  
2 - val\_mae: 0.3578  
Epoch 28/100  
50/50  0s 4ms/step - loss: 0.3228 - mae: 0.3330 - val\_loss: 0.299  
6 - val\_mae: 0.3702  
Epoch 29/100  
50/50  0s 4ms/step - loss: 0.3642 - mae: 0.4041 - val\_loss: 0.289  
9 - val\_mae: 0.3331  
Epoch 30/100  
50/50  0s 4ms/step - loss: 0.2993 - mae: 0.3506 - val\_loss: 0.325  
2 - val\_mae: 0.4135  
Epoch 31/100  
50/50  0s 4ms/step - loss: 0.3575 - mae: 0.3647 - val\_loss: 0.286  
8 - val\_mae: 0.3486  
Epoch 32/100  
50/50  0s 5ms/step - loss: 0.2161 - mae: 0.2835 - val\_loss: 0.264  
6 - val\_mae: 0.3201  
Epoch 33/100  
50/50  0s 4ms/step - loss: 0.2969 - mae: 0.3333 - val\_loss: 0.304  
4 - val\_mae: 0.4240  
Epoch 34/100  
50/50  0s 4ms/step - loss: 0.3064 - mae: 0.3456 - val\_loss: 0.262  
9 - val\_mae: 0.3269  
Epoch 35/100  
50/50  0s 4ms/step - loss: 0.3397 - mae: 0.3290 - val\_loss: 0.295  
8 - val\_mae: 0.3912  
Epoch 36/100  
50/50  0s 3ms/step - loss: 0.2547 - mae: 0.3155 - val\_loss: 0.243  
6 - val\_mae: 0.3056  
Epoch 37/100  
50/50  0s 3ms/step - loss: 0.2935 - mae: 0.3245 - val\_loss: 0.241  
3 - val\_mae: 0.3022  
Epoch 38/100  
50/50  0s 4ms/step - loss: 0.2536 - mae: 0.2982 - val\_loss: 0.231  
6 - val\_mae: 0.2981  
Epoch 39/100  
50/50  0s 4ms/step - loss: 0.2147 - mae: 0.2588 - val\_loss: 0.223  
2 - val\_mae: 0.3010  
Epoch 40/100  
50/50  0s 3ms/step - loss: 0.2477 - mae: 0.3094 - val\_loss: 0.224  
6 - val\_mae: 0.2922

Epoch 41/100  
50/50  0s 4ms/step - loss: 0.2134 - mae: 0.2669 - val\_loss: 0.214  
1 - val\_mae: 0.3211  
Epoch 42/100  
50/50  0s 3ms/step - loss: 0.2150 - mae: 0.2951 - val\_loss: 0.210  
5 - val\_mae: 0.2843  
Epoch 43/100  
50/50  0s 4ms/step - loss: 0.2589 - mae: 0.3075 - val\_loss: 0.224  
3 - val\_mae: 0.3530  
Epoch 44/100  
50/50  0s 4ms/step - loss: 0.2007 - mae: 0.2975 - val\_loss: 0.203  
2 - val\_mae: 0.3273  
Epoch 45/100  
50/50  0s 4ms/step - loss: 0.1784 - mae: 0.2598 - val\_loss: 0.190  
7 - val\_mae: 0.2780  
Epoch 46/100  
50/50  0s 4ms/step - loss: 0.1896 - mae: 0.2866 - val\_loss: 0.198  
3 - val\_mae: 0.3178  
Epoch 47/100  
50/50  0s 4ms/step - loss: 0.2290 - mae: 0.3428 - val\_loss: 0.183  
9 - val\_mae: 0.2811  
Epoch 48/100  
50/50  0s 4ms/step - loss: 0.1778 - mae: 0.2618 - val\_loss: 0.181  
5 - val\_mae: 0.2768  
Epoch 49/100  
50/50  0s 4ms/step - loss: 0.1710 - mae: 0.2521 - val\_loss: 0.179  
5 - val\_mae: 0.2797  
Epoch 50/100  
50/50  0s 4ms/step - loss: 0.2041 - mae: 0.2888 - val\_loss: 0.200  
9 - val\_mae: 0.3521  
Epoch 51/100  
50/50  0s 4ms/step - loss: 0.1939 - mae: 0.2929 - val\_loss: 0.154  
6 - val\_mae: 0.2533  
Epoch 52/100  
50/50  0s 4ms/step - loss: 0.1773 - mae: 0.2517 - val\_loss: 0.152  
6 - val\_mae: 0.2599  
Epoch 53/100  
50/50  0s 4ms/step - loss: 0.1998 - mae: 0.2623 - val\_loss: 0.149  
9 - val\_mae: 0.2474  
Epoch 54/100  
50/50  0s 4ms/step - loss: 0.1572 - mae: 0.2616 - val\_loss: 0.261  
5 - val\_mae: 0.4347  
Epoch 55/100  
50/50  0s 3ms/step - loss: 0.1919 - mae: 0.3201 - val\_loss: 0.145  
0 - val\_mae: 0.2532  
Epoch 56/100  
50/50  0s 3ms/step - loss: 0.1298 - mae: 0.2366 - val\_loss: 0.136  
7 - val\_mae: 0.2366  
Epoch 57/100  
50/50  0s 3ms/step - loss: 0.1641 - mae: 0.2479 - val\_loss: 0.171  
6 - val\_mae: 0.2918  
Epoch 58/100  
50/50  0s 4ms/step - loss: 0.1698 - mae: 0.2667 - val\_loss: 0.140  
7 - val\_mae: 0.2426  
Epoch 59/100  
50/50  0s 4ms/step - loss: 0.1494 - mae: 0.2398 - val\_loss: 0.131  
4 - val\_mae: 0.2416  
Epoch 60/100  
50/50  0s 4ms/step - loss: 0.1176 - mae: 0.2243 - val\_loss: 0.147  
5 - val\_mae: 0.3089

Epoch 61/100  
50/50  0s 4ms/step - loss: 0.1578 - mae: 0.2962 - val\_loss: 0.135  
3 - val\_mae: 0.2516  
Epoch 62/100  
50/50  0s 4ms/step - loss: 0.1302 - mae: 0.2515 - val\_loss: 0.139  
3 - val\_mae: 0.2613  
Epoch 63/100  
50/50  0s 4ms/step - loss: 0.1391 - mae: 0.2595 - val\_loss: 0.140  
6 - val\_mae: 0.2710  
Epoch 64/100  
50/50  0s 4ms/step - loss: 0.1953 - mae: 0.3268 - val\_loss: 0.114  
8 - val\_mae: 0.2265  
Epoch 65/100  
50/50  0s 4ms/step - loss: 0.1459 - mae: 0.2490 - val\_loss: 0.140  
3 - val\_mae: 0.2635  
Epoch 66/100  
50/50  0s 4ms/step - loss: 0.1399 - mae: 0.2539 - val\_loss: 0.123  
2 - val\_mae: 0.2694  
Epoch 67/100  
50/50  0s 4ms/step - loss: 0.1488 - mae: 0.2712 - val\_loss: 0.122  
4 - val\_mae: 0.2605  
Epoch 68/100  
50/50  0s 4ms/step - loss: 0.1728 - mae: 0.3027 - val\_loss: 0.109  
3 - val\_mae: 0.2324  
Epoch 69/100  
50/50  0s 4ms/step - loss: 0.1764 - mae: 0.3157 - val\_loss: 0.145  
7 - val\_mae: 0.2896  
Epoch 70/100  
50/50  0s 4ms/step - loss: 0.1598 - mae: 0.2907 - val\_loss: 0.110  
4 - val\_mae: 0.2440  
Epoch 71/100  
50/50  0s 4ms/step - loss: 0.1310 - mae: 0.2464 - val\_loss: 0.115  
3 - val\_mae: 0.2588  
Epoch 72/100  
50/50  0s 4ms/step - loss: 0.1218 - mae: 0.2493 - val\_loss: 0.096  
2 - val\_mae: 0.2338  
Epoch 73/100  
50/50  0s 5ms/step - loss: 0.1411 - mae: 0.2717 - val\_loss: 0.094  
8 - val\_mae: 0.2312  
Epoch 74/100  
50/50  0s 4ms/step - loss: 0.1205 - mae: 0.2538 - val\_loss: 0.151  
5 - val\_mae: 0.2946  
Epoch 75/100  
50/50  0s 4ms/step - loss: 0.1244 - mae: 0.2691 - val\_loss: 0.254  
1 - val\_mae: 0.3370  
Epoch 76/100  
50/50  0s 4ms/step - loss: 0.1811 - mae: 0.3147 - val\_loss: 0.102  
4 - val\_mae: 0.2427  
Epoch 77/100  
50/50  0s 4ms/step - loss: 0.1078 - mae: 0.2332 - val\_loss: 0.088  
6 - val\_mae: 0.2197  
Epoch 78/100  
50/50  0s 4ms/step - loss: 0.1196 - mae: 0.2552 - val\_loss: 0.097  
2 - val\_mae: 0.2271  
Epoch 79/100  
50/50  0s 4ms/step - loss: 0.1097 - mae: 0.2291 - val\_loss: 0.110  
3 - val\_mae: 0.2568  
Epoch 80/100  
50/50  0s 4ms/step - loss: 0.1095 - mae: 0.2490 - val\_loss: 0.105  
2 - val\_mae: 0.2444



Epoch 81/100  
50/50  0s 5ms/step - loss: 0.1212 - mae: 0.2555 - val\_loss: 0.1113 - val\_mae: 0.2726  
Epoch 82/100  
50/50  0s 6ms/step - loss: 0.1509 - mae: 0.3061 - val\_loss: 0.1289 - val\_mae: 0.2901  
Epoch 83/100  
50/50  0s 6ms/step - loss: 0.1571 - mae: 0.3031 - val\_loss: 0.1210 - val\_mae: 0.2765  
Epoch 84/100  
50/50  0s 4ms/step - loss: 0.1423 - mae: 0.2811 - val\_loss: 0.1081 - val\_mae: 0.2698  
Epoch 85/100  
50/50  0s 4ms/step - loss: 0.1135 - mae: 0.2500 - val\_loss: 0.1191 - val\_mae: 0.2803  
Epoch 86/100  
50/50  0s 5ms/step - loss: 0.1293 - mae: 0.2641 - val\_loss: 0.2610 - val\_mae: 0.4219  
Epoch 87/100  
50/50  0s 4ms/step - loss: 0.1813 - mae: 0.3282 - val\_loss: 0.0946 - val\_mae: 0.2428  
Epoch 88/100  
50/50  0s 4ms/step - loss: 0.1150 - mae: 0.2569 - val\_loss: 0.1326 - val\_mae: 0.2805  
Epoch 89/100  
50/50  0s 5ms/step - loss: 0.1352 - mae: 0.2773 - val\_loss: 0.1201 - val\_mae: 0.2662  
Epoch 90/100  
50/50  0s 4ms/step - loss: 0.1342 - mae: 0.2810 - val\_loss: 0.1919 - val\_mae: 0.3536  
Epoch 91/100  
50/50  0s 4ms/step - loss: 0.1495 - mae: 0.2907 - val\_loss: 0.0745 - val\_mae: 0.2046  
Epoch 92/100  
50/50  0s 4ms/step - loss: 0.1202 - mae: 0.2583 - val\_loss: 0.0819 - val\_mae: 0.2171  
Epoch 93/100  
50/50  0s 4ms/step - loss: 0.1094 - mae: 0.2401 - val\_loss: 0.1615 - val\_mae: 0.3259  
Epoch 94/100  
50/50  0s 4ms/step - loss: 0.1420 - mae: 0.2954 - val\_loss: 0.2668 - val\_mae: 0.4322  
Epoch 95/100  
50/50  0s 4ms/step - loss: 0.1617 - mae: 0.3131 - val\_loss: 0.1186 - val\_mae: 0.2805  
Epoch 96/100  
50/50  0s 4ms/step - loss: 0.1083 - mae: 0.2476 - val\_loss: 0.0845 - val\_mae: 0.2255  
Epoch 97/100  
50/50  0s 5ms/step - loss: 0.1018 - mae: 0.2403 - val\_loss: 0.1122 - val\_mae: 0.2754  
Epoch 98/100  
50/50  0s 4ms/step - loss: 0.1346 - mae: 0.2884 - val\_loss: 0.0986 - val\_mae: 0.2389  
Epoch 99/100  
50/50  0s 4ms/step - loss: 0.1237 - mae: 0.2614 - val\_loss: 0.1544 - val\_mae: 0.3158  
Epoch 100/100  
50/50  0s 4ms/step - loss: 0.1068 - mae: 0.2453 - val\_loss: 0.0717 - val\_mae: 0.2083

7/7 ————— 0s 6ms/step - loss: 0.0733 - mae: 0.2088  
 Deep Learning Model - Loss: 0.07, MAE: 0.21

```
In [10]: # Predict QoE for test data
y_pred_dl = deep_model.predict(X_test)

# Convert predictions to DataFrame
predictions_df = pd.DataFrame({'Actual QoE': y_test.values, 'Predicted QoE': y_pred_dl})

# Display first few rows
print(predictions_df.head())

# Plot actual vs predicted QoE scores
plt.figure(figsize=(8, 4))
plt.scatter(y_test, y_pred_dl, color='blue', alpha=0.5)
plt.plot([1, 5], [1, 5], '--', color='red') # Ideal line
plt.xlabel("Actual QoE Score")
plt.ylabel("Predicted QoE Score")
plt.title("Actual vs Predicted QoE Scores (Deep Learning Model)")
plt.grid()
plt.show()
```

7/7 ————— 0s 14ms/step

	Actual QoE	Predicted QoE
0	1.000000	0.957789
1	1.000000	1.234877
2	1.000000	1.274469
3	2.801309	1.980495
4	1.000000	1.127310

```
-----
NameError                                Traceback (most recent call last)
Cell In[10], line 11
      8 print(predictions_df.head())
     10 # Plot actual vs predicted QoE scores
----> 11 plt.figure(figsize=(8, 4))
     12 plt.scatter(y_test, y_pred_dl, color='blue', alpha=0.5)
     13 plt.plot([1, 5], [1, 5], '--', color='red') # Ideal line

NameError: name 'plt' is not defined
```

```
In [14]: # Predict QoE for test data
y_pred_dl = deep_model.predict(X_test)

# Convert predictions to DataFrame
predictions_df = pd.DataFrame({'Actual QoE': y_test.values, 'Predicted QoE': y_pred_dl})

# Display first few rows
print(predictions_df.head())

# Plot actual vs predicted QoE scores
plt.figure(figsize=(8, 4))
plt.scatter(y_test, y_pred_dl, color='blue', alpha=0.5)
plt.plot([1, 5], [1, 5], '--', color='red') # Ideal line
plt.xlabel("Actual QoE Score")
plt.ylabel("Predicted QoE Score")
plt.title("Actual vs Predicted QoE Scores (Deep Learning Model)")
plt.grid()
plt.show()
```

7/7 ————— 0s 8ms/step

	Actual QoE	Predicted QoE
0	1.000000	0.957789
1	1.000000	1.234877
2	1.000000	1.274469
3	2.801309	1.980495
4	1.000000	1.127310

```
-----
NameError                                Traceback (most recent call last)
Cell In[14], line 11
      8 print(predictions_df.head())
     10 # Plot actual vs predicted QoE scores
----> 11 plt.figure(figsize=(8, 4))
     12 plt.scatter(y_test, y_pred_d1, color='blue', alpha=0.5)
     13 plt.plot([1, 5], [1, 5], '--', color='red') # Ideal line

NameError: name 'plt' is not defined
```

In [16]: `import matplotlib.pyplot as plt`

In [18]: `import matplotlib.pyplot as plt`

In [20]: `import matplotlib.pyplot as plt # Ensure this is imported before using plt`

```
# Predict QoE for test data
y_pred_d1 = deep_model.predict(X_test)

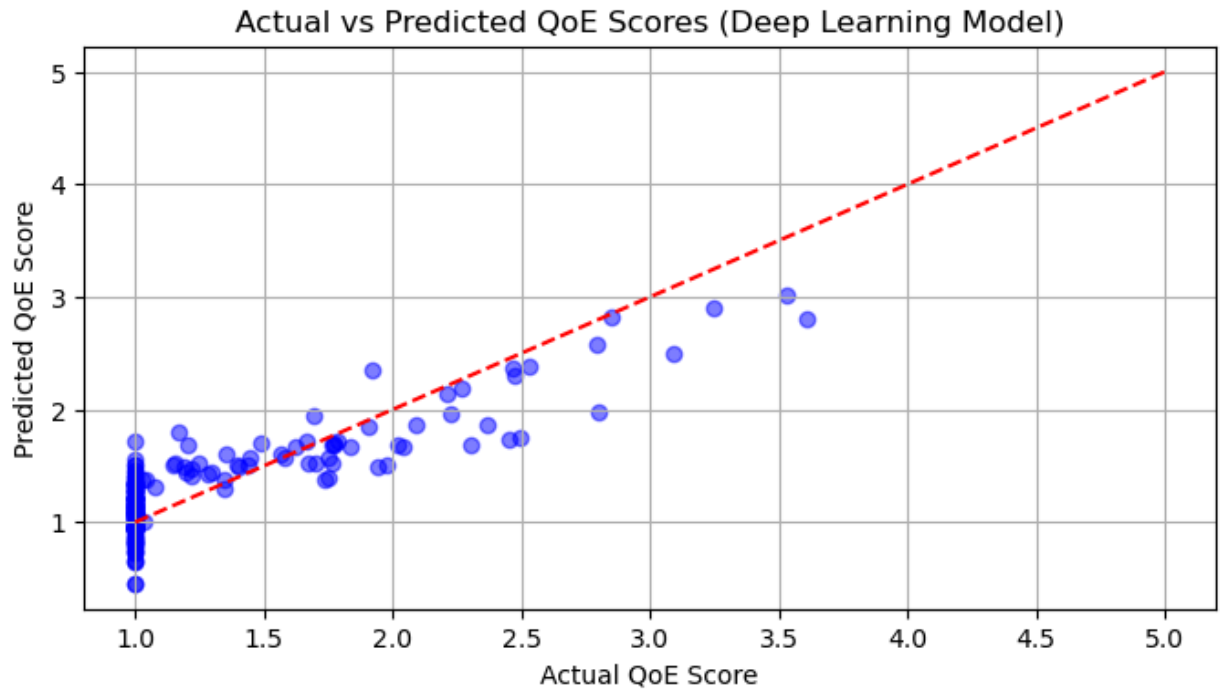
# Convert predictions to DataFrame
predictions_df = pd.DataFrame({'Actual QoE': y_test.values, 'Predicted QoE': y_pred_d1})

# Display first few rows
print(predictions_df.head())

# Plot actual vs predicted QoE scores
plt.figure(figsize=(8, 4))
plt.scatter(y_test, y_pred_d1, color='blue', alpha=0.5)
plt.plot([1, 5], [1, 5], '--', color='red') # Ideal line
plt.xlabel("Actual QoE Score")
plt.ylabel("Predicted QoE Score")
plt.title("Actual vs Predicted QoE Scores (Deep Learning Model)")
plt.grid()
plt.show()
```

7/7 ————— 0s 7ms/step

	Actual QoE	Predicted QoE
0	1.000000	0.957789
1	1.000000	1.234877
2	1.000000	1.274469
3	2.801309	1.980495
4	1.000000	1.127310



In [24]: `import matplotlib.pyplot as plt # Ensure matplotlib is imported`

```
# Compare all models
models = ["Linear Regression", "Random Forest", "Deep Learning"]
mse_values = [mse_lr, mse_rf, mse_dl]
r2_values = [r2_lr, r2_rf, r2_dl]

# Plot MSE Comparison
plt.figure(figsize=(8, 4))
plt.bar(models, mse_values, color=['blue', 'green', 'red'])
plt.xlabel("Models")
plt.ylabel("Mean Squared Error (MSE)")
plt.title("Model Comparison - MSE")
plt.show()

# Plot R2 Score Comparison
plt.figure(figsize=(8, 4))
plt.bar(models, r2_values, color=['blue', 'green', 'red'])
plt.xlabel("Models")
plt.ylabel("R2 Score")
plt.title("Model Comparison - R2 Score")
plt.show()
```

```
-----
NameError                                Traceback (most recent call last)
Cell In[24], line 5
      3 # Compare all models
      4 models = ["Linear Regression", "Random Forest", "Deep Learning"]
----> 5 mse_values = [mse_lr, mse_rf, mse_dl]
      6 r2_values = [r2_lr, r2_rf, r2_dl]
      8 # Plot MSE Comparison

NameError: name 'mse_lr' is not defined
```

In [30]: `from sklearn.linear_model import LinearRegression # Import Linear Regression model`  
`from sklearn.ensemble import RandomForestRegressor # Import Random Forest`

```
from sklearn.metrics import mean_squared_error, r2_score # Import evaluation metrics
from sklearn.model_selection import train_test_split # Import data splitting function
```

```
In [32]: import matplotlib.pyplot as plt # Ensure matplotlib is imported

# Ensure required libraries are imported
from sklearn.linear_model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.model_selection import train_test_split

# Recreate input-output data if not already defined
X = df[['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)', 'Bandwidth (Mbps)']]
y = df['QoE Score']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Train Linear Regression Model
lr_model = LinearRegression()
lr_model.fit(X_train, y_train)
y_pred_lr = lr_model.predict(X_test)
mse_lr = mean_squared_error(y_test, y_pred_lr)
r2_lr = r2_score(y_test, y_pred_lr)

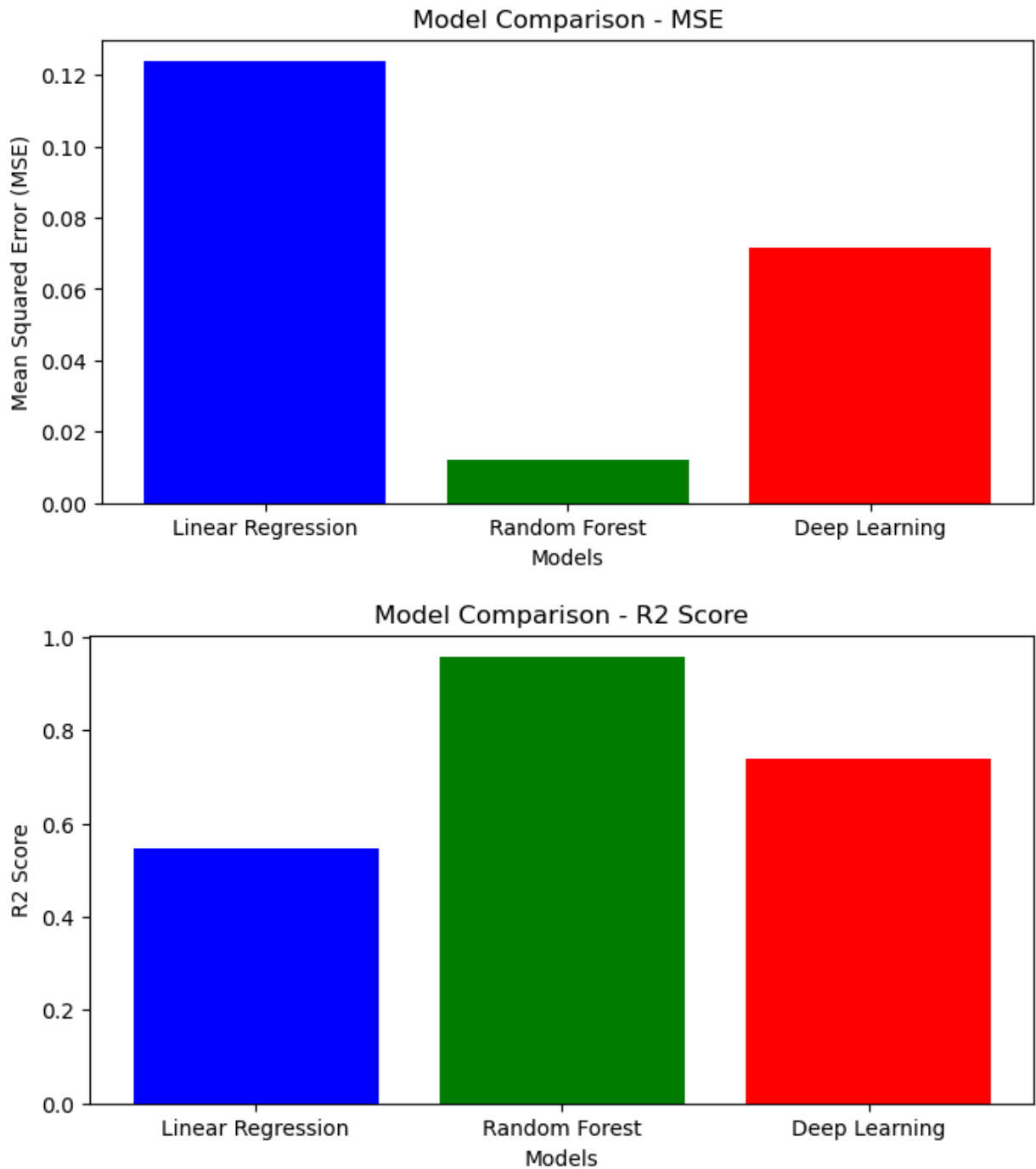
# Train Random Forest Model
rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)
y_pred_rf = rf_model.predict(X_test)
mse_rf = mean_squared_error(y_test, y_pred_rf)
r2_rf = r2_score(y_test, y_pred_rf)

# Train Deep Learning Model if not already done
if 'mse_dl' not in globals():
    y_pred_dl = deep_model.predict(X_test)
    mse_dl = mean_squared_error(y_test, y_pred_dl)
    r2_dl = r2_score(y_test, y_pred_dl)

# Compare all models
models = ["Linear Regression", "Random Forest", "Deep Learning"]
mse_values = [mse_lr, mse_rf, mse_dl]
r2_values = [r2_lr, r2_rf, r2_dl]

# Plot MSE Comparison
plt.figure(figsize=(8, 4))
plt.bar(models, mse_values, color=['blue', 'green', 'red'])
plt.xlabel("Models")
plt.ylabel("Mean Squared Error (MSE)")
plt.title("Model Comparison - MSE")
plt.show()

# Plot R2 Score Comparison
plt.figure(figsize=(8, 4))
plt.bar(models, r2_values, color=['blue', 'green', 'red'])
plt.xlabel("Models")
plt.ylabel("R2 Score")
plt.title("Model Comparison - R2 Score")
plt.show()
```



```
In [34]: def predict_qoe_deep(latency, packet_loss, jitter, bandwidth):
    input_data = pd.DataFrame([[latency, packet_loss, jitter, bandwidth]],
                               columns=['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)'])
    qoe_score = deep_model.predict(input_data)[0][0]
    return round(qoe_score, 2)

# Test the function with sample network parameters
latency = 50
packet_loss = 1.5
jitter = 15
bandwidth = 50
predicted_qoe_dl = predict_qoe_deep(latency, packet_loss, jitter, bandwidth)
print(f"Predicted QoE Score (Deep Learning): {predicted_qoe_dl}/5")
```

1/1 ————— 0s 69ms/step  
 Predicted QoE Score (Deep Learning): 2.4700000286102295/5

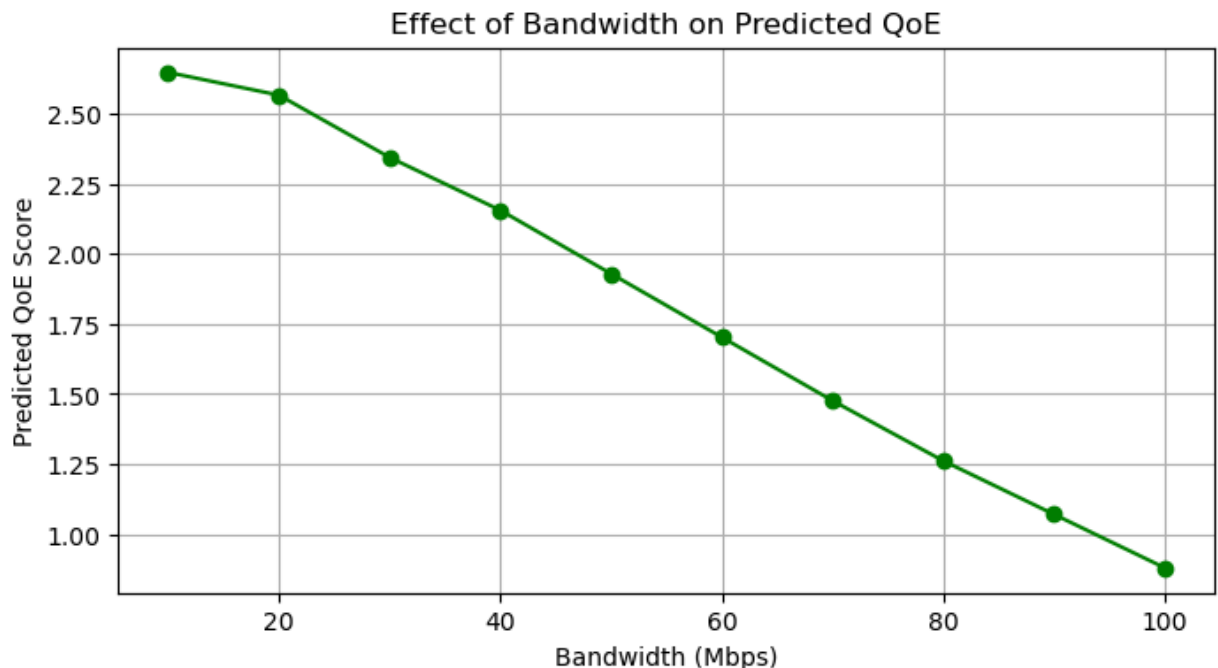
```
In [36]: # Generate multiple network conditions
network_conditions = pd.DataFrame({
    'Latency (ms)': np.linspace(20, 200, 10),
    'Packet Loss (%)': np.linspace(0, 3, 10),
    'Jitter (ms)': np.linspace(5, 50, 10),
    'Bandwidth (Mbps)': np.linspace(10, 100, 10)
})

# Predict QoE scores using deep Learning model
network_conditions['Predicted QoE'] = deep_model.predict(network_conditions)

# Find best condition
optimal_condition = network_conditions.loc[network_conditions['Predicted QoE'].idxmax()]
print("Optimal Network Parameters for Best QoE:")
print(optimal_condition)

# Visualizing the effect of bandwidth on QoE
plt.figure(figsize=(8, 4))
plt.plot(network_conditions['Bandwidth (Mbps)'], network_conditions['Predicted QoE'],
plt.title("Effect of Bandwidth on Predicted QoE")
plt.xlabel("Bandwidth (Mbps)")
plt.ylabel("Predicted QoE Score")
plt.grid()
plt.show()
```

1/1 ————— 0s 63ms/step  
Optimal Network Parameters for Best QoE:  
Latency (ms) 20.000  
Packet Loss (%) 0.000  
Jitter (ms) 5.000  
Bandwidth (Mbps) 10.000  
Predicted QoE 2.648  
Name: 0, dtype: float64



```
In [38]: # Function to suggest network parameter adjustments
def optimize_network(latency, packet_loss, jitter, bandwidth):
    current_qoe = predict_qoe_deep(latency, packet_loss, jitter, bandwidth)
```

```



    if current_qoe < 3:
        if latency > 100: latency -= 20
        if packet_loss > 2: packet_loss -= 0.5
        if jitter > 20: jitter -= 5
        if bandwidth < 50: bandwidth += 10

    optimized_qoe = predict_qoe_deep(latency, packet_loss, jitter, bandwidth)

    return {
        "Original QoE": current_qoe,
        "Optimized QoE": optimized_qoe,
        "Suggested Latency": latency,
        "Suggested Packet Loss": packet_loss,
        "Suggested Jitter": jitter,
        "Suggested Bandwidth": bandwidth
    }

# Example usage
network_status = optimize_network(120, 3, 25, 30)
print("Network Optimization Result:", network_status)

```

1/1  0s 171ms/step  
1/1  0s 49ms/step

Network Optimization Result: {'Original QoE': 1.44, 'Optimized QoE': 1.69, 'Suggested Latency': 100, 'Suggested Packet Loss': 2.5, 'Suggested Jitter': 20, 'Suggested Bandwidth': 40}

```

In [40]: # Generate summary table for different network conditions
results = []
for latency in [50, 100, 150, 200]:
    for loss in [1, 2, 3]:
        for jitter in [10, 20, 30]:
            for bandwidth in [20, 50, 80]:
                qoe = predict_qoe_deep(latency, loss, jitter, bandwidth)
                results.append([latency, loss, jitter, bandwidth, qoe])

# Convert to DataFrame
summary_df = pd.DataFrame(results, columns=['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)', 'Bandwidth (Mbps)', 'QoE'])

# Display summary
print("Summary of QoE Predictions under Different Conditions:")
print(summary_df)

# Save as CSV for documentation
summary_df.to_csv("QoE_Prediction_Report.csv", index=False)

# Visualize QoE distribution
plt.figure(figsize=(8, 4))
sns.histplot(summary_df['Predicted QoE'], bins=20, kde=True, color='purple')
plt.title("Distribution of Predicted QoE Scores")
plt.xlabel("Predicted QoE Score")
plt.ylabel("Frequency")
plt.show()

```



1/1	0s	52ms/step
1/1	0s	49ms/step
1/1	0s	43ms/step
1/1	0s	45ms/step
1/1	0s	47ms/step
1/1	0s	43ms/step
1/1	0s	50ms/step
1/1	0s	43ms/step
1/1	0s	46ms/step
1/1	0s	51ms/step
1/1	0s	42ms/step
1/1	0s	44ms/step
1/1	0s	41ms/step
1/1	0s	46ms/step
1/1	0s	46ms/step
1/1	0s	46ms/step
1/1	0s	46ms/step
1/1	0s	44ms/step
1/1	0s	46ms/step
1/1	0s	43ms/step
1/1	0s	42ms/step
1/1	0s	47ms/step
1/1	0s	46ms/step
1/1	0s	51ms/step
1/1	0s	54ms/step
1/1	0s	61ms/step
1/1	0s	68ms/step
1/1	0s	62ms/step
1/1	0s	53ms/step
1/1	0s	59ms/step
1/1	0s	67ms/step
1/1	0s	71ms/step
1/1	0s	51ms/step
1/1	0s	50ms/step
1/1	0s	48ms/step
1/1	0s	57ms/step
1/1	0s	86ms/step
1/1	0s	51ms/step
1/1	0s	49ms/step
1/1	0s	52ms/step
1/1	0s	62ms/step
1/1	0s	53ms/step
1/1	0s	50ms/step
1/1	0s	55ms/step
1/1	0s	46ms/step
1/1	0s	54ms/step
1/1	0s	56ms/step
1/1	0s	51ms/step
1/1	0s	46ms/step
1/1	0s	52ms/step
1/1	0s	54ms/step
1/1	0s	50ms/step
1/1	0s	56ms/step
1/1	0s	58ms/step
1/1	0s	52ms/step
1/1	0s	66ms/step
1/1	0s	55ms/step
1/1	0s	52ms/step
1/1	0s	48ms/step
1/1	0s	48ms/step

1/1		0s	53ms/step
1/1		0s	54ms/step
1/1		0s	54ms/step
1/1		0s	51ms/step
1/1		0s	49ms/step
1/1		0s	48ms/step
1/1		0s	52ms/step
1/1		0s	49ms/step
1/1		0s	48ms/step
1/1		0s	53ms/step
1/1		0s	51ms/step
1/1		0s	44ms/step
1/1		0s	57ms/step
1/1		0s	51ms/step
1/1		0s	55ms/step
1/1		0s	51ms/step
1/1		0s	51ms/step
1/1		0s	66ms/step
1/1		0s	90ms/step
1/1		0s	46ms/step
1/1		0s	44ms/step
1/1		0s	53ms/step
1/1		0s	50ms/step
1/1		0s	48ms/step
1/1		0s	53ms/step
1/1		0s	52ms/step
1/1		0s	70ms/step
1/1		0s	60ms/step
1/1		0s	54ms/step
1/1		0s	55ms/step
1/1		0s	61ms/step
1/1		0s	58ms/step
1/1		0s	63ms/step
1/1		0s	61ms/step
1/1		0s	63ms/step
1/1		0s	56ms/step
1/1		0s	48ms/step
1/1		0s	52ms/step
1/1		0s	80ms/step
1/1		0s	57ms/step
1/1		0s	54ms/step
1/1		0s	59ms/step
1/1		0s	63ms/step
1/1		0s	55ms/step
1/1		0s	60ms/step
1/1		0s	51ms/step
1/1		0s	59ms/step
1/1		0s	52ms/step

Summary of QoE Predictions under Different Conditions:

	Latency (ms)	Packet Loss (%)	Jitter (ms)	Bandwidth (Mbps)	\
0	50	1	10	20	
1	50	1	10	50	
2	50	1	10	80	
3	50	1	20	20	
4	50	1	20	50	
..	...	...	...	...	
103	200	3	20	50	
104	200	3	20	80	
105	200	3	30	20	
106	200	3	30	50	

107	200	3	30	80
-----	-----	---	----	----

	Predicted QoE
0	2.27
1	2.81
2	2.95
3	2.17
4	2.55
..	...
103	1.21
104	1.36
105	1.16
106	1.12
107	1.30

[108 rows x 5 columns]

```

-----
NameError                                Traceback (most recent call last)
Cell In[40], line 22
    20 # Visualize QoE distribution
    21 plt.figure(figsize=(8, 4))
--> 22 sns.histplot(summary_df['Predicted QoE'], bins=20, kde=True, color='purple')
    23 plt.title("Distribution of Predicted QoE Scores")
    24 plt.xlabel("Predicted QoE Score")

NameError: name 'sns' is not defined
<Figure size 800x400 with 0 Axes>

```

```
In [42]: import seaborn as sns # Import Seaborn for visualization
```

```
In [44]: # Generate summary table for different network conditions
results = []
for latency in [50, 100, 150, 200]:
    for loss in [1, 2, 3]:
        for jitter in [10, 20, 30]:
            for bandwidth in [20, 50, 80]:
                qoe = predict_qoe_deep(latency, loss, jitter, bandwidth)
                results.append([latency, loss, jitter, bandwidth, qoe])

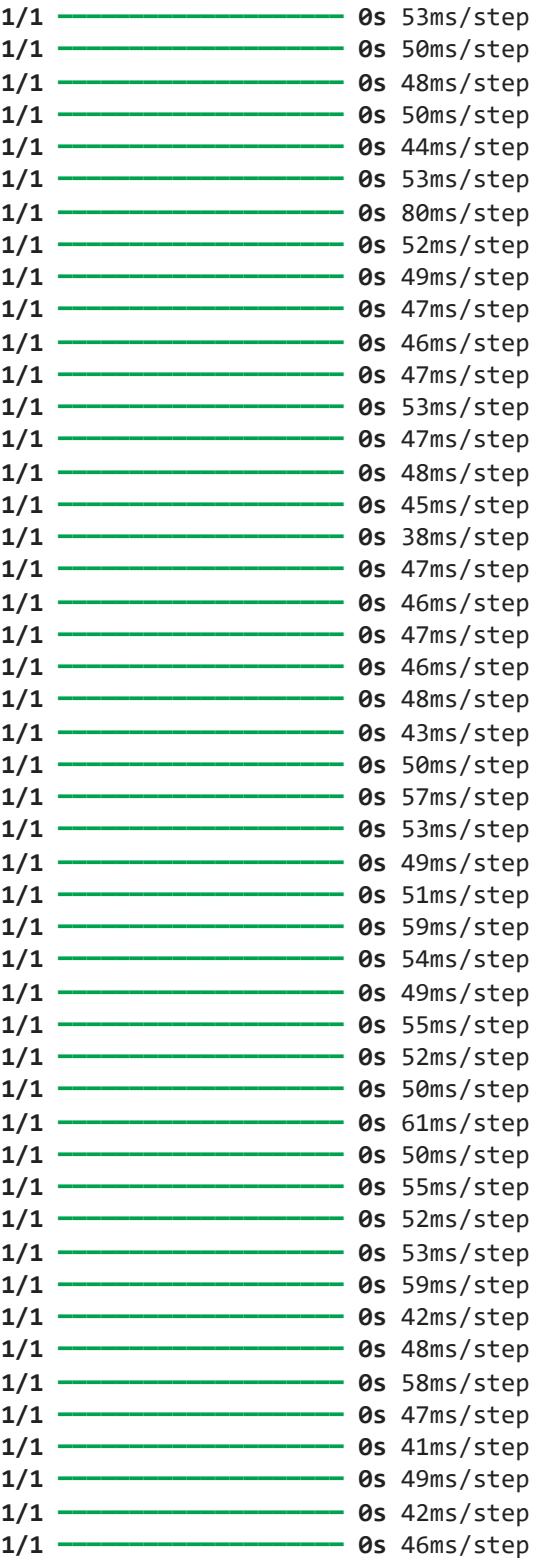
# Convert to DataFrame
summary_df = pd.DataFrame(results, columns=['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)', 'Bandwidth (Mbps)', 'Predicted QoE'])

# Display summary
print("Summary of QoE Predictions under Different Conditions:")
print(summary_df)

# Save as CSV for documentation
summary_df.to_csv("QoE_Prediction_Report.csv", index=False)

# Visualize QoE distribution
plt.figure(figsize=(8, 4))
sns.histplot(summary_df['Predicted QoE'], bins=20, kde=True, color='purple')
plt.title("Distribution of Predicted QoE Scores")
plt.xlabel("Predicted QoE Score")
plt.ylabel("Frequency")
plt.show()
```

1/1		0s	71ms/step
1/1		0s	50ms/step
1/1		0s	45ms/step
1/1		0s	45ms/step
1/1		0s	45ms/step
1/1		0s	50ms/step
1/1		0s	53ms/step
1/1		0s	42ms/step
1/1		0s	45ms/step
1/1		0s	45ms/step
1/1		0s	46ms/step
1/1		0s	45ms/step
1/1		0s	43ms/step
1/1		0s	47ms/step
1/1		0s	43ms/step
1/1		0s	50ms/step
1/1		0s	52ms/step
1/1		0s	47ms/step
1/1		0s	58ms/step
1/1		0s	54ms/step
1/1		0s	55ms/step
1/1		0s	53ms/step
1/1		0s	54ms/step
1/1		0s	51ms/step
1/1		0s	48ms/step
1/1		0s	47ms/step
1/1		0s	44ms/step
1/1		0s	45ms/step
1/1		0s	43ms/step
1/1		0s	43ms/step
1/1		0s	42ms/step
1/1		0s	46ms/step
1/1		0s	53ms/step
1/1		0s	45ms/step
1/1		0s	46ms/step
1/1		0s	48ms/step
1/1		0s	48ms/step
1/1		0s	53ms/step
1/1		0s	46ms/step
1/1		0s	49ms/step
1/1		0s	51ms/step
1/1		0s	45ms/step
1/1		0s	44ms/step
1/1		0s	54ms/step
1/1		0s	45ms/step
1/1		0s	42ms/step
1/1		0s	51ms/step
1/1		0s	48ms/step
1/1		0s	51ms/step
1/1		0s	58ms/step
1/1		0s	50ms/step
1/1		0s	52ms/step
1/1		0s	54ms/step
1/1		0s	57ms/step
1/1		0s	52ms/step
1/1		0s	50ms/step
1/1		0s	45ms/step
1/1		0s	49ms/step
1/1		0s	57ms/step
1/1		0s	50ms/step

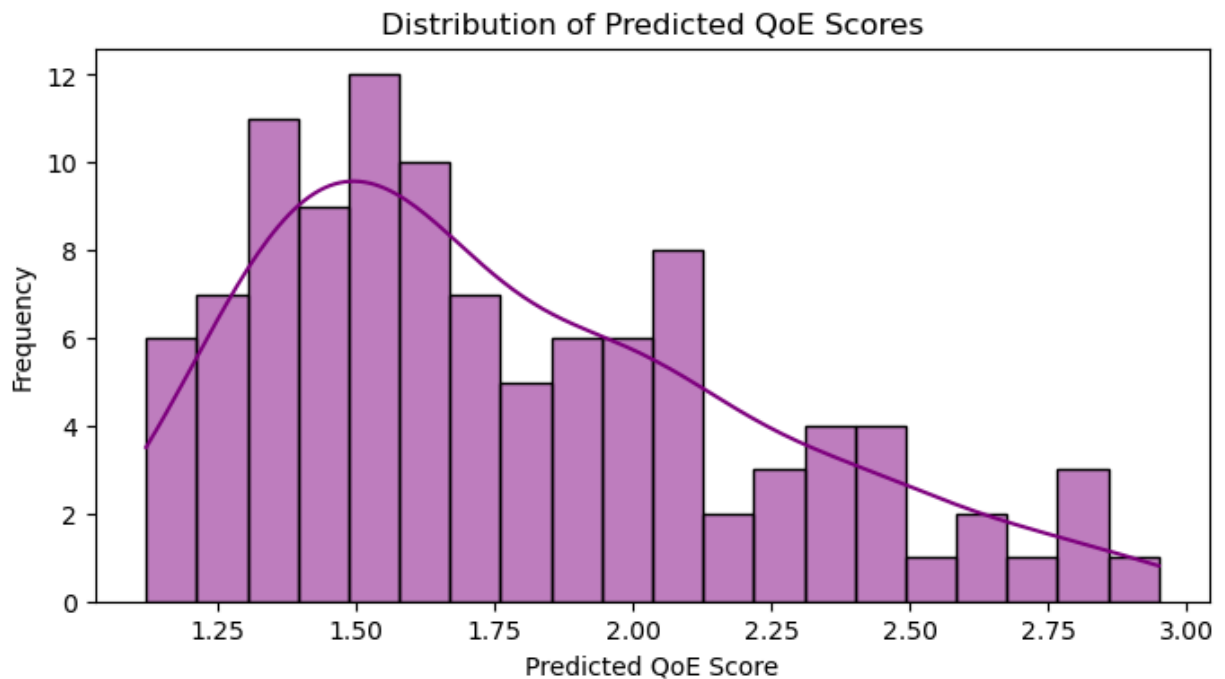


Summary of QoE Predictions under Different Conditions:

	Latency (ms)	Packet Loss (%)	Jitter (ms)	Bandwidth (Mbps)	\
0	50	1	10	20	
1	50	1	10	50	
2	50	1	10	80	
3	50	1	20	20	
4	50	1	20	50	
..	...	...	...	...	
103	200	3	20	50	
104	200	3	20	80	
105	200	3	30	20	
106	200	3	30	50	

107	200	3	30	80
Predicted QoE				
0	2.27			
1	2.81			
2	2.95			
3	2.17			
4	2.55			
..	...			
103	1.21			
104	1.36			
105	1.16			
106	1.12			
107	1.30			

[108 rows x 5 columns]



```
In [46]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import random
from matplotlib.animation import FuncAnimation

# Initialize figure
fig, ax = plt.subplots(figsize=(8, 4))
x_data, y_data = [], []
line, = ax.plot([], [], 'bo-', markersize=5)

# Set labels
ax.set_xlim(0, 100)
ax.set_ylim(1, 5)
ax.set_xlabel("Time")
ax.set_ylabel("QoE Score")
ax.set_title("Real-time QoE Monitoring Simulation")

# Function to simulate real-time data
def update(frame):
```

```

global x_data, y_data

latency = random.randint(20, 300) # Simulating Latency (ms)
jitter = random.randint(5, 100) # Simulating Jitter (ms)
packet_loss = random.uniform(0, 5) # Simulating Packet Loss (%)
bandwidth = random.uniform(5, 100) # Simulating Bandwidth (Mbps)

# Calculate QoE dynamically
qoe_score = 5 - (latency / 80 + packet_loss / 2 + jitter / 30)
qoe_score = np.clip(qoe_score, 1, 5) # Ensure QoE stays between 1-5

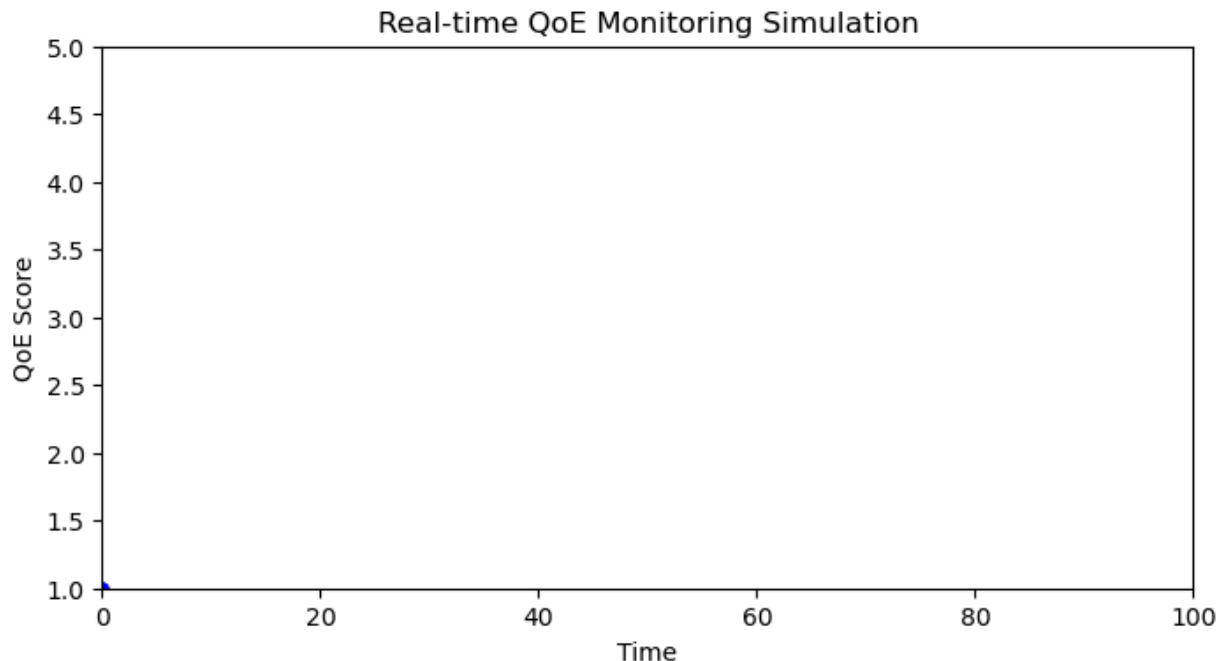
x_data.append(frame)
y_data.append(qoe_score)

# Keep only last 100 points
if len(x_data) > 100:
    x_data.pop(0)
    y_data.pop(0)

line.set_data(x_data, y_data)
return line,

# Run animation
ani = FuncAnimation(fig, update, frames=range(100), interval=500, blit=True)
plt.show()

```



```

In [48]: def simulate_qoe():
    latency = float(input("Enter Latency (ms): "))
    packet_loss = float(input("Enter Packet Loss (%): "))
    jitter = float(input("Enter Jitter (ms): "))
    bandwidth = float(input("Enter Bandwidth (Mbps): "))

    # Predict QoE using the trained Random Forest model
    input_data = pd.DataFrame([[latency, packet_loss, jitter, bandwidth]],
                               columns=['Latency (ms)', 'Packet Loss (%)', 'Jitter (ms)'])
    predicted_qoe = rf_model.predict(input_data)[0]

    print(f"Predicted QoE Score: {round(predicted_qoe, 2)} / 5")

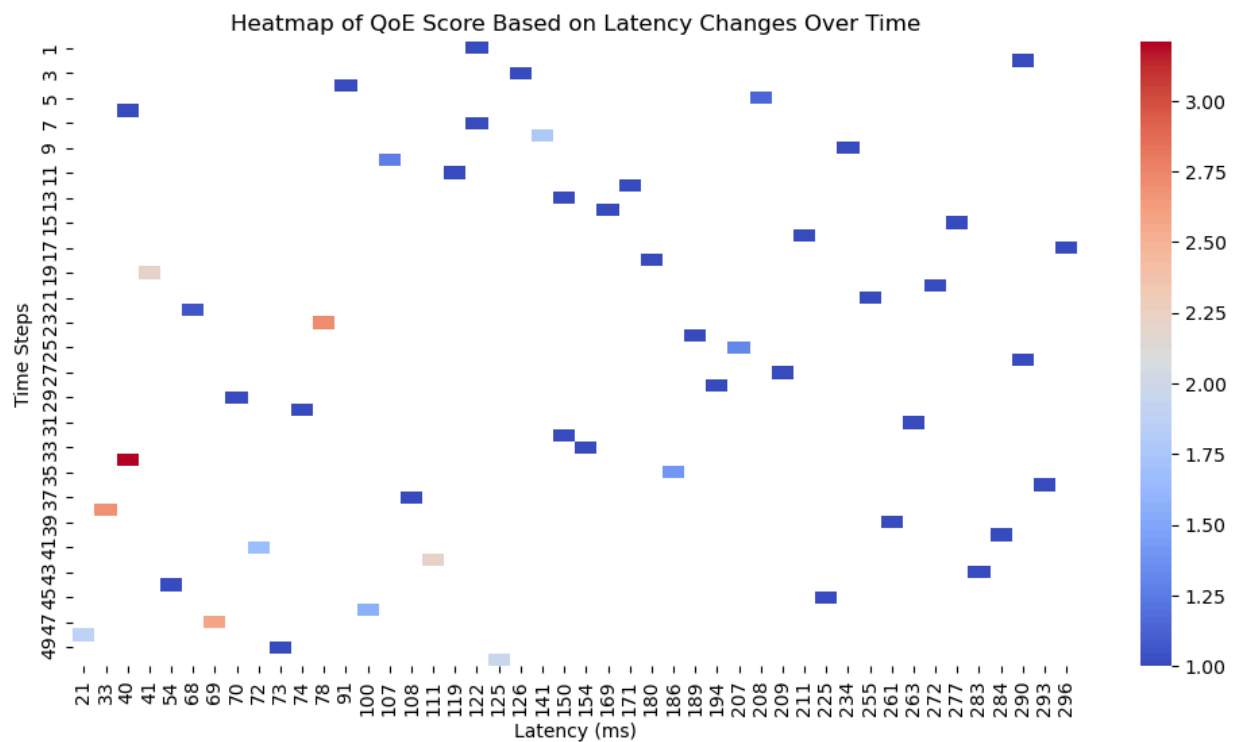
```

```
# Run user input simulation  
simulate_qoe()
```

Predicted QoE Score: 1.16 / 5

```
In [52]: import seaborn as sns  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
  
# Generate simulated data  
np.random.seed(42)  
time_steps = np.arange(1, 51) # Simulating 50 time steps  
latency = np.random.randint(20, 300, 50)  
packet_loss = np.random.uniform(0, 5, 50)  
jitter = np.random.randint(5, 100, 50)  
bandwidth = np.random.uniform(5, 100, 50)  
  
# Compute QoE Scores  
qoe_scores = 5 - (latency / 80 + packet_loss / 2 + jitter / 30)  
qoe_scores = np.clip(qoe_scores, 1, 5) # Ensure QoE stays between 1-5  
  
# Create a DataFrame  
df_sim = pd.DataFrame({'Time': time_steps, 'Latency': latency, 'QoE Score': qoe_scores})  
  
# Use pivot_table instead of pivot to handle duplicate "Latency" values at the same time  
df_pivot = df_sim.pivot_table(index="Time", columns="Latency", values="QoE Score", aggfunc="mean")  
  
# Plot Heatmap  
plt.figure(figsize=(12, 6))  
sns.heatmap(df_pivot, cmap="coolwarm", annot=False)  
plt.title("Heatmap of QoE Score Based on Latency Changes Over Time")  
plt.xlabel("Latency (ms)")  
plt.ylabel("Time Steps")  
plt.show()
```





```
In [54]: import time

def real_time_qoe_prediction():
    print("Starting Real-Time QoE Simulation...\n")

    for i in range(10): # Simulate for 10 time steps
        latency = np.random.randint(20, 300)
        jitter = np.random.randint(5, 100)
        packet_loss = np.random.uniform(0, 5)
        bandwidth = np.random.uniform(5, 100)

        # Predict QoE using deep learning model
        input_data = np.array([[latency, packet_loss, jitter, bandwidth]])
        predicted_qoe = deep_model.predict(input_data)[0][0]

        print(f"Time Step {i+1}: Latency={latency}ms, Jitter={jitter}ms, Packet Loss={packet_loss}%, Bandwidth={bandwidth}Mbps, Predicted QoE={predicted_qoe}")
        time.sleep(1) # Simulating real-time delay

    # Run real-time QoE prediction
    real_time_qoe_prediction()
```

Starting Real-Time QoE Simulation...

1/1  0s 53ms/step

Time Step 1: Latency=131ms, Jitter=64ms, Packet Loss=1.30%, Bandwidth=99.64 Mbps → Predicted QoE: 1.24

1/1  0s 45ms/step

Time Step 2: Latency=273ms, Jitter=16ms, Packet Loss=2.79%, Bandwidth=88.85 Mbps → Predicted QoE: 1.08

1/1  0s 43ms/step

Time Step 3: Latency=28ms, Jitter=23ms, Packet Loss=3.63%, Bandwidth=90.23 Mbps → Predicted QoE: 2.29

1/1  0s 39ms/step

Time Step 4: Latency=171ms, Jitter=58ms, Packet Loss=3.90%, Bandwidth=65.99 Mbps → Predicted QoE: 1.17

1/1  0s 53ms/step

Time Step 5: Latency=94ms, Jitter=76ms, Packet Loss=3.93%, Bandwidth=68.55 Mbps → Predicted QoE: 1.14

1/1  0s 47ms/step

Time Step 6: Latency=273ms, Jitter=93ms, Packet Loss=3.32%, Bandwidth=5.48 Mbps → Predicted QoE: 0.93

1/1  0s 44ms/step

Time Step 7: Latency=213ms, Jitter=58ms, Packet Loss=2.24%, Bandwidth=99.47 Mbps → Predicted QoE: 0.98

1/1  0s 44ms/step

Time Step 8: Latency=180ms, Jitter=72ms, Packet Loss=1.19%, Bandwidth=35.91 Mbps → Predicted QoE: 1.38

1/1  0s 44ms/step

Time Step 9: Latency=154ms, Jitter=71ms, Packet Loss=3.29%, Bandwidth=58.99 Mbps → Predicted QoE: 1.11

1/1  0s 49ms/step

Time Step 10: Latency=41ms, Jitter=34ms, Packet Loss=4.87%, Bandwidth=42.34 Mbps → Predicted QoE: 0.80

```
In [4]: import numpy as np # Import NumPy
import pandas as pd # Import Pandas (since you're using DataFrame)
```

```
In [6]: import numpy as np
import pandas as pd

# Simulating realistic network traffic conditions
time_steps = np.arange(1000) # Simulating over 1000 time steps
data = {
    'Time': time_steps,
    'Latency (ms)': np.random.normal(100, 50, 1000), # Mean=100, Std=50
    'Packet Loss (%)': np.random.normal(1.5, 1, 1000),
    'Jitter (ms)': np.random.normal(20, 10, 1000),
    'Bandwidth (Mbps)': np.random.normal(50, 20, 1000)
}

df = pd.DataFrame(data)

# Generate QoE Score using a weighted formula
df['QoE Score'] = 5 - (df['Latency (ms)'] / 80 + df['Packet Loss (%)'] / 2 + df['Jitter (ms)'] / 10)
df['QoE Score'] = np.clip(df['QoE Score'], 1, 5) # Ensure QoE stays between 1 and 5

# Display first few rows
df.head()
```

Out[6]:

	Time	Latency (ms)	Packet Loss (%)	Jitter (ms)	Bandwidth (Mbps)	QoE Score
0	0	108.460495	1.188576	33.306166	67.552600	1.939750
1	1	19.659397	1.686803	31.860837	55.707161	2.848828
2	2	91.255006	1.748453	17.421558	51.544013	2.404367
3	3	54.207655	2.555885	20.366863	66.699487	2.365566
4	4	104.315674	2.146345	15.259815	38.300453	2.114221