

```

from ucimlrepo import fetch_ucirepo
import pandas as pd
# fetch dataset
statlog_heart = fetch_ucirepo(id=145)

```

```

# data (as pandas dataframes)
X = statlog_heart.data.features
y = statlog_heart.data.targets

```

```
print(X.head())
```

```

➡      age  sex  chest-pain  rest-bp  serum-chol  fasting-blood-sugar  \
0  70.0  1.0         4.0    130.0        322.0             0.0
1  67.0  0.0         3.0    115.0        564.0             0.0
2  57.0  1.0         2.0    124.0        261.0             0.0
3  64.0  1.0         4.0    128.0        263.0             0.0
4  74.0  0.0         2.0    120.0        269.0             0.0

      electrocardiographic  max-heart-rate  angina  oldpeak  slope  \
0                2.0        109.0      0.0      2.4    2.0
1                2.0        160.0      0.0      1.6    2.0
2                0.0        141.0      0.0      0.3    1.0
3                0.0        105.0      1.0      0.2    2.0
4                2.0        121.0      1.0      0.2    1.0

      major-vessels  thal
0                3.0  3.0
1                0.0  7.0
2                0.0  7.0
3                1.0  7.0
4                1.0  3.0

```

```

# metadata
print(statlog_heart.metadata)

```

```

# variable information
print(statlog_heart.variables)

```

```

X_encoded = pd.get_dummies(X, drop_first=True)
print(X_encoded.head())

```

```

➡      age  sex  chest-pain  rest-bp  serum-chol  fasting-blood-sugar  \
0  70.0  1.0         4.0    130.0        322.0             0.0
1  67.0  0.0         3.0    115.0        564.0             0.0
2  57.0  1.0         2.0    124.0        261.0             0.0
3  64.0  1.0         4.0    128.0        263.0             0.0
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4                2.0        121.0      1.0      0.2    1.0

```

1	2.0	160.0	0.0	1.6	2.0
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3	0.0	105.0	1.0	0.2	2.0
4	2.0	121.0	1.0	0.2	1.0

	major-vessels	thal
0	3.0	3.0
1	0.0	7.0
2	0.0	7.0
3	1.0	7.0
4	1.0	3.0

```
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
```

```
encoder = OneHotEncoder(drop='first') # drop='first' avoids multicollinearity
categorical_columns = ['chest-pain', 'electrocardiographic', 'thal']
column_transformer = ColumnTransformer(transformers=[('encoder', encoder, categorical_col
X_encoded = column_transformer.fit_transform(X)
X_encoded_df = pd.DataFrame(X_encoded)
print(X_encoded_df.head())
```



	0	1	2	3	4	5	6	7	8	9	10	11	12	\
0	0.0	0.0	1.0	0.0	1.0	0.0	0.0	70.0	1.0	130.0	322.0	0.0	109.0	
1	0.0	1.0	0.0	0.0	1.0	0.0	1.0	67.0	0.0	115.0	564.0	0.0	160.0	
2	1.0	0.0	0.0	0.0	0.0	0.0	1.0	57.0	1.0	124.0	261.0	0.0	141.0	
3	0.0	0.0	1.0	0.0	0.0	0.0	1.0	64.0	1.0	128.0	263.0	0.0	105.0	
4	1.0	0.0	0.0	0.0	1.0	0.0	0.0	74.0	0.0	120.0	269.0	0.0	121.0	

	13	14	15	16
0	0.0	2.4	2.0	3.0
1	0.0	1.6	2.0	0.0
2	0.0	0.3	1.0	0.0
3	1.0	0.2	2.0	1.0
4	1.0	0.2	1.0	1.0

```
from sklearn.decomposition import PCA
```

```
print(f"Original feature shape: {X.shape}")
print(f"Target shape: {y.shape}")
```

```
# Apply PCA to reduce features from 13 to 8
pca = PCA(n_components=8)
X_reduced = pca.fit_transform(X_encoded)
```

```
X_pca_df = pd.DataFrame(X_reduced, columns=[f'PC{i+1}' for i in range(8)])
print(f"Reduced feature shape: {X_reduced.shape}")
```

```
Original feature shape: (270, 13)
Target shape: (270, 1)
Reduced feature shape: (270, 8)
```

```
from qiskit.circuit.library import ZZFeatureMap, PauliFeatureMap
def get_pauli(feature_dimension = 8, reps = 2):
    return PauliFeatureMap(feature_dimension=feature_dimension, paulis=['Z', 'YY'], reps=

def get_zzfeaturemap(feature_dimension = 8, reps = 2, entanglement = 'full'):
    return ZZFeatureMap(feature_dimension=feature_dimension, reps=reps, entanglement=enta
```

```
from sklearn.model_selection import train_test_split
# split into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X_reduced, y, test_size=0.20, random_
# split into training and validation sets
X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.125, rand
```

```
from qiskit import QuantumCircuit
import numpy as np
```

```
def custom_ansatz_two(num_qubits, layers, entanglement='pairwise', rotation_gates=None, e
"""
```

Constructs a Two-Local ansatz circuit with fixed theta values.

Parameters:

num_qubits (int): The number of qubits in the circuit.
 layers (int): The number of rotation and entanglement layers.
 entanglement (str): The strategy for entangling qubits ('linear', 'circular', 'pa
 rotation_gates (list): List of rotation gates to use (e.g., ['ry', 'rz']).
 entanglement_gates (list): List of entanglement gates to use (e.g., ['cx']).

Returns:

QuantumCircuit: The constructed quantum circuit.

"""

```
circuit = QuantumCircuit(num_qubits)
```

```
# Define default rotation and entanglement gates if none provided
```

```
if rotation_gates is None:
```

```
    rotation_gates = ['ry']
```

```
if entanglement_gates is None:
```

```
    entanglement_gates = ['cx'] # Default entanglement gates
```

```

entanglement_gates = ['cx'] # Default entanglement gate

for layer in range(layers):
    # Apply rotation layer
    for qubit in range(num_qubits):
        for gate in rotation_gates:

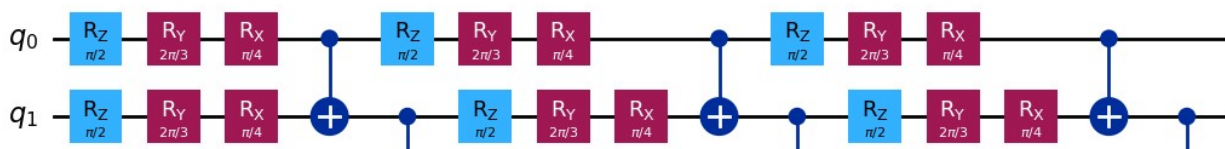
            if gate == 'ry':
                circuit.ry((np.pi/3)*2, qubit)
            elif gate == 'rz':
                circuit.rz(np.pi/2, qubit)
            elif gate == 'rx':
                circuit.rx(np.pi/4, qubit)

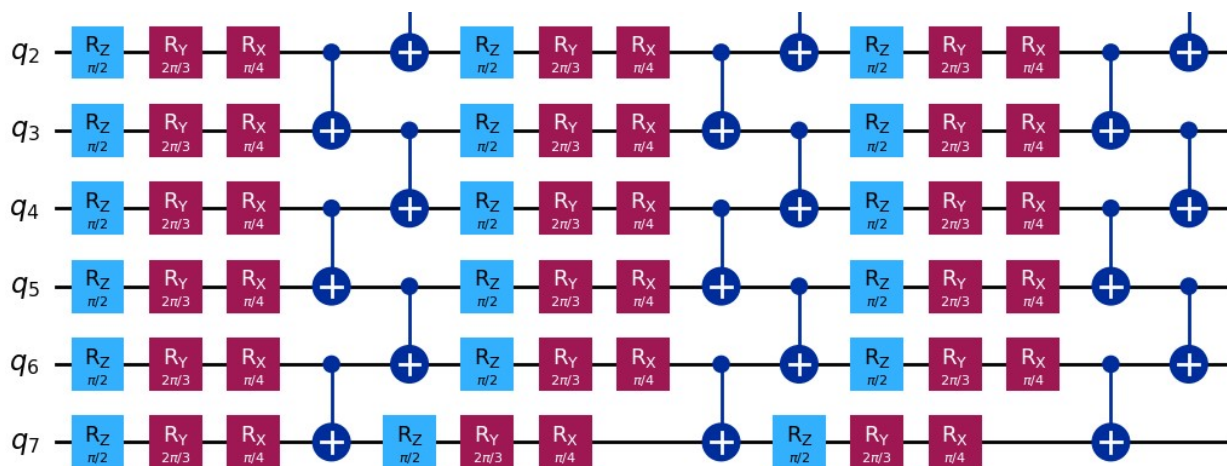
    if entanglement == 'linear':
        for i in range(num_qubits - 1):
            for gate in entanglement_gates:
                if gate == 'cx':
                    circuit.cx(i, i + 1)
    elif entanglement == 'circular':
        for i in range(num_qubits):
            for gate in entanglement_gates:
                if gate == 'cx':
                    circuit.cx(i, (i + 1) % num_qubits) # Circular entanglement
    elif entanglement == 'pairwise':
        for i in range(0, num_qubits - 1, 2): # Even indices
            for gate in entanglement_gates:
                if gate == 'cx':
                    circuit.cx(i, i + 1) # Entangle qubit i with i + 1
        for i in range(1, num_qubits - 1, 2): # Odd indices
            for gate in entanglement_gates:
                if gate == 'cx':
                    circuit.cx(i, i + 1) # Entangle qubit i with i + 1

return circuit

# Example:
num_qubits = 8
layers = 3
circuit = custom_ansatz_two(num_qubits, layers, rotation_gates=['rz','ry','rx'], entangle
circuit.draw(output='mpl')

```





```
from qiskit_machine_learning.algorithms.classifiers import VQC
from qiskit_algorithms.optimizers import COBYLA
from qiskit_algorithms.optimizers import SPSA
```

```
from qiskit_aer import QasmSimulator
from qiskit.primitives import BackendSampler
```

```
zz = get_zzfeaturemap()
backend = QasmSimulator()
```

```
import pandas as pd
from qiskit_machine_learning.algorithms.classifiers import VQC
from qiskit_algorithms.optimizers import COBYLA
from qiskit_aer import QasmSimulator
from qiskit.primitives import BackendSampler
from sklearn.metrics import accuracy_score, f1_score, precision_score, recall_score
import time
```

```
entanglement_strategies = ['linear', 'circular', 'pairwise']
backend = QasmSimulator()
```

```
detailed_results = []
accuracy_results = []

# Loop over each entanglement strategy
for entanglement in entanglement_strategies:
    print(f"Testing entanglement strategy: {entanglement}")

    # Create ansatz with current entanglement strategy
    ansatz = custom_ansatz_two(num_qubits=8, layers=6, entanglement=entanglement, rotation=0)
    ansatz.draw(output='mpl')

    vqc = VQC(
        feature_map=get_zzfeaturemap(),
        ansatz=ansatz,
        optimizer=COBYLA(maxiter=500),
        sampler=BackendSampler(backend=backend)
    )

    start = time.time()
    vqc.fit(X_train, y_train.to_numpy())
    end = time.time()

    y_val_pred = vqc.predict(X_val)
    val_acc = accuracy_score(y_val, y_val_pred)
    val_f1 = f1_score(y_val, y_val_pred, average="weighted")
    val_precision = precision_score(y_val, y_val_pred, average="weighted")
    val_recall = recall_score(y_val, y_val_pred, average="weighted")

    y_test_pred = vqc.predict(X_test)
    test_acc = accuracy_score(y_test, y_test_pred)
    test_f1 = f1_score(y_test, y_test_pred, average="weighted")
    test_precision = precision_score(y_test, y_test_pred, average="weighted")
    test_recall = recall_score(y_test, y_test_pred, average="weighted")

    detailed_results.append({
        'Entanglement': entanglement,
        'Validation Accuracy': val_acc,
        'Validation F1 Score': val_f1,
        'Validation Precision': val_precision,
        'Validation Recall': val_recall,
        'Test Accuracy': test_acc,
        'Test F1 Score': test_f1,
        'Test Precision': test_precision,
        'Test Recall': test_recall,
        'Training Time (s)': end - start
    })

    accuracy_results.append({
        'Entanglement': entanglement,
        'Validation Accuracy': val_acc,
```

```

        'Test Accuracy': test_acc
    })

detailed_results_df = pd.DataFrame(detailed_results)

accuracy_results_df = pd.DataFrame(accuracy_results)

print("\nFinal Accuracy Results:")
print(accuracy_results_df)

print("\nDetailed Metrics Results:")
print(detailed_results_df, end=' ')

##GATES USED:
#2RY Gates, 2pi/3

```

```

Testing entanglement strategy: linear
Testing entanglement strategy: circular
Testing entanglement strategy: pairwise

```

Final Accuracy Results:

	Entanglement	Validation Accuracy	Test Accuracy
0	linear	0.481481	0.574074
1	circular	0.592593	0.481481
2	pairwise	0.629630	0.611111

Detailed Metrics Results:

	Entanglement	Validation Accuracy	Validation F1 Score \
0	linear	0.481481	0.478632
1	circular	0.592593	0.590307
2	pairwise	0.629630	0.630647

	Validation Precision	Validation Recall	Test Accuracy	Test F1 Score \
0	0.497475	0.481481	0.574074	0.574805
1	0.589646	0.592593	0.481481	0.482912
2	0.638584	0.629630	0.611111	0.607499

	Test Precision	Test Recall	Training Time (s)
0	0.585441	0.574074	2.179157
1	0.485653	0.481481	1.622357
2	0.639250	0.611111	1.565358

```

from qiskit import QuantumCircuit
import numpy as np

```

```

def custom_pauli(num_qubits): #full entanglement
    circuit = QuantumCircuit(num_qubits)
    for layer in range(3):
        for qubit in range(num_qubits):

```

```

circuit.ry(theta=np.pi/2, qubit=qubit)
circuit.rx(theta=np.pi,qubit=qubit)
circuit.rz(np.pi*3,qubit=qubit)

```

```

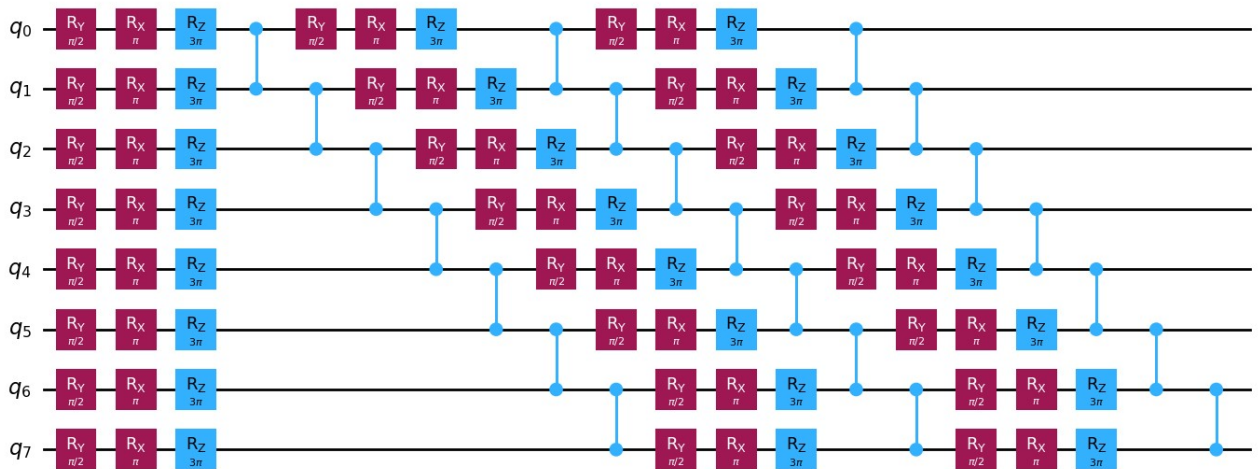
# Entangle all qubits, full entanglement
for i in range(num_qubits - 1):
    circuit.cz(i, i + 1)
return circuit

```

```

ansatz_pauli = custom_pauli(num_qubits=8)
ansatz_pauli.draw(output='mpl')

```



```

from sklearn.metrics import accuracy_score, f1_score, precision_score, recall_score
import time

```

```

pauli = get_pauli()

```



```
vqc_pauli = VQC(feature_map=pauli, ansatz=ansatz_pauli, optimizer=COBYLA(maxiter=500), sampl

start = time.time()
vqc_pauli.fit(X_train, y_train.to_numpy())
end = time.time()

y_train_pred = vqc_pauli.predict(X_train)

y_val_pred = vqc_pauli.predict(X_val)

val_acc = accuracy_score(y_val, y_val_pred)
val_f1 = f1_score(y_val, y_val_pred, average="weighted")
val_precision = precision_score(y_val, y_val_pred, average="weighted")
val_recall = recall_score(y_val, y_val_pred, average="weighted")

print(f"Validation results: Accuracy = {val_acc:.2f}, F1 Score = {val_f1:.2f}, Precision =

y_test_pred = vqc_pauli.predict(X_test)

test_acc = accuracy_score(y_test, y_test_pred)
test_f1 = f1_score(y_test, y_test_pred, average="weighted")
test_precision = precision_score(y_test, y_test_pred, average="weighted")
test_recall = recall_score(y_test, y_test_pred, average="weighted")

print(f"Test results: Accuracy = {test_acc:.2f}, F1 Score = {test_f1:.2f}, Precision = {te:

print(f"Fitted vqc_pauli with training time = {end - start:.2f} seconds")
```

```
Validation results: Accuracy = 0.81, F1 Score = 0.81, Precision = 0.82, Recall = 0.81
Test results: Accuracy = 0.57, F1 Score = 0.58, Precision = 0.58, Recall = 0.57
Fitted vqc_pauli with training time = 1.76 seconds
```

