Machine Learning-Derived Temperature Trajectories Identify Coagulation Phenotypes in Sepsis

```
1.preprocess temperature data
def preprocess temperature data(df):
  #标准化:以每个患者前3次体温测量的均值为基线
  df['temp normalized'] = df.groupby('patient id')['temperature'].transform(
    lambda x: x / x.head(3).mean()
  )
  #立方样条插值处理缺失值
  df['temp interpolated'] = df.groupby('patient id')['temp normalized'].transform(
    lambda x: x.interpolate(method='cubic')
 )
2.cluster temperature trajectories)
def cluster temperature trajectories(trajectories):
  #DTW距离矩阵计算
  for i in range(n):
    alignment = dtw(trajectories[i], trajectories[i])
    dtw matrix[i,j] = alignment.distance
  #GMM聚类 + BIC选择最优簇数
  gmm = GaussianMixture(n components=optimal clusters)
  labels = gmm.fit predict(dtw matrix)
3.analyze coagulation by phenotype
results[phenotype] = {
  'platelet count': subset['platelet'].mean(), #血小板计数
  'pt ratio': subset['pt ratio'].mean(), #凝血酶原时间比值
  'd dimer': subset['d dimer'].mean(),
                                       # D-二聚体水平
  'mortality rate': subset['mortality'].mean() # 死亡率
```

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Full data:
import numpy as np
import pandas as pd
from sklearn.mixture import GaussianMixture
from dtw import dtw
from sklearn.metrics import silhouette score
from sklearn.preprocessing import StandardScaler
def preprocess temperature data(df):
  """Preprocess temperature data as described in Methods section"""
  # Normalize to baseline (mean of first 3 measurements)
  df['temp normalized'] = df.groupby('patient id')['temperature'].transform(
    lambda x: x / x.head(3).mean()
  )
  # Handle missing data with cubic spline interpolation
  df['temp interpolated'] = df.groupby('patient id')['temp normalized'].transform(
    lambda x: x.interpolate(method='cubic')
  )
  return df
def cluster temperature trajectories(trajectories, max clusters=5):
  """Cluster temperature trajectories using DTW and GMM"""
  # Calculate DTW distance matrix
  n = len(trajectories)
  dtw matrix = np.zeros((n, n))
  for i in range(n):
    for j in range(i+1, n):
      alignment = dtw(trajectories[i], trajectories[j])
      dtw matrix[i,j] = alignment.distance
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dtw matrix[j,i] = alignment.distance
  # Find optimal cluster number using BIC
  bic scores = []
  silhouette scores = []
  for n clusters in range(2, max clusters+1):
    gmm = GaussianMixture(n_clusters=n_clusters)
    labels = gmm.fit predict(dtw matrix)
    bic_scores.append(gmm.bic(dtw_matrix))
    silhouette scores.append(silhouette score(dtw matrix, labels))
  optimal clusters = np.argmin(bic scores) + 2 # +2 because range starts at 2
  # Final clustering with optimal clusters
  gmm = GaussianMixture(n components=optimal clusters)
  cluster labels = gmm.fit predict(dtw matrix)
  return cluster labels, optimal clusters, silhouette scores[optimal clusters-2]
def analyze coagulation by phenotype(data, labels):
  """Analyze coagulation parameters by phenotype"""
  results = \{\}
  for phenotype in np.unique(labels):
    subset = data[labels == phenotype]
    results[phenotype] = {
       'platelet count': subset['platelet'].mean(),
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'platelet slope': subset['platelet slope'].mean(),

'pt ratio': subset['pt ratio'].mean(),

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'd dimer': subset['d dimer'].mean(),
      'mortality rate': subset['mortality'].mean()
    }
  return pd.DataFrame.from dict(results, orient='index')
# Example usage:
if name == " main ":
  # Load sample data (replace with actual data loading)
  data = pd.read csv('sepsis data.csv')
  # Preprocess temperature data
  processed data = preprocess temperature data(data)
  # Extract trajectories (assuming data is in long format)
          trajectories = processed_data.groupby('patient_id')
['temp interpolated'].apply(np.array).values
  # Cluster trajectories
  labels, n clusters, silhouette = cluster temperature trajectories(trajectories)
  print(f"Identified {n_clusters} phenotypes with silhouette score {silhouette:.2f}")
  # Analyze coagulation parameters by phenotype
  phenotype analysis = analyze coagulation by phenotype(data, labels)
  print(phenotype analysis)
```