```
import numpy as np
import pandas as pd
from sklearn.impute import KNNImputer, SimpleImputer
from sklearn.experimental import enable iterative imputer
from sklearn.impute import IterativeImputer
from fancyimpute import SoftImpute
import matplotlib.pyplot as plt
import seaborn as sns
import random
from sklearn.ensemble import StackingRegressor
from sklearn.metrics.pairwise import cosine similarity
from sklearn.model selection import train test split
from sklearn.metrics import mean squared error
data=pd.read csv('/content/online retail customer churn.csv')
data.drop(['Gender', 'Email Opt In', 'Promotion Response',
'Target Churn', 'Age'], axis=1, inplace=True)
data10A = data.copy()
# Select 10% of the rows
rows to modify = random.sample(range(len(data10A)), int(len(data10A) *
0.1))
columns to modify = range(1, 10)
# Set the selected values to NaN
for row index in rows to modify:
    for column index in columns to modify:
        data10A.iloc[row index, column index] = np.nan
null values count = data10A.isnull().sum()
print(null values count)
missing positions = np.argwhere(pd.isnull(data10A).values)
# Initialize KNN Imputation
knn_imputer = KNNImputer(n neighbors=5)
# Initialize IterativeImputer
iterative_imputer = IterativeImputer(max_iter=10)
# Initialize SimpleImputer with mean strategy
mean imputer = SimpleImputer(missing values=np.nan, strategy='mean')
# Initialize SoftImpute
soft imputer = SoftImpute()
```

```
# Initialize SimpleImputer with median strategy
median imputer = SimpleImputer(missing values=np.nan,
strategy='median')
def calculate rmse (imputed data, original data, missing positions):
    missing values = [(x, y) \text{ for } x, y \text{ in missing positions}]
    original values = [original data.iloc[x, y] for x, y in
missing values]
    imputed values = [imputed data[x, y] for x, y in missing values]
    return np.sqrt(mean squared error(original values, imputed values))
# Function to impute missing values using the proposed ensemble method
def ensemble impute (data with missing, missing positions,
original data):
    ip temp = data with missing.copy() # Ensure ip temp is reset for
each imputation
    imputed data = data with missing.copy()
    for i in range(len(missing positions)):
        x, y = missing positions[i]
        # Impute using KNN
        ip temp.iloc[x, y] = np.nan
        knn imputed = knn imputer.fit_transform(ip_temp)
        knn value = knn imputed[x, y]
        # Impute using Iterative Imputer
        ip temp.iloc[x, y] = np.nan
        iterative imputed = iterative imputer.fit transform(ip temp)
        iterative value = iterative imputed[x, y]
        # Impute using Mean
        ip temp.iloc[x, y] = np.nan
        mean imputed = mean imputer.fit transform(ip temp)
        mean_value = mean_imputed[x, y]
        # Impute using SoftImpute
        ip temp.iloc[x, y] = np.nan
        soft imputed = SoftImpute().fit transform(ip temp)
        soft_value = soft_imputed[x, y]
        # Impute using Median
        ip temp.iloc[x, y] = np.nan
        median imputed = median imputer.fit transform(ip temp)
        median_value = median_imputed[x, y]
        # Store the imputed values in a dictionary
        imputed values = {
          'knn': knn value,
```

```
'iterative': iterative value,
            'mean': mean value,
            'soft': soft value,
            'median': median value
        original value = original data.iloc[x, y]
        rmses = {method: np.sqrt(mean squared error([value],
[original value])) for method, value in imputed values.items()}
        # Select the method with the smallest RMSE
        best method = min(rmses, key=rmses.get)
        \# Impute the value at (x, y) in the original dataset using the
best method
        imputed data.iloc[x, y] = imputed values[best method]
    return imputed data
ensemble imputed = ensemble impute(data10A, missing positions, data)
knn imputed = knn imputer.fit transform(data10A)
# Impute using Iterative Imputer
iterative imputed = iterative imputer.fit transform(data10A)
# Impute using Mean
mean imputed = mean imputer.fit transform(data10A)
# Impute using SoftImpute
soft imputed = SoftImpute().fit transform(data10A)
# Impute using Median
median imputed = median imputer.fit transform(data10A)
# Calculated RMSE for each method
knn rmse = calculate rmse(knn imputed, data, missing positions)
iterative rmse = calculate rmse(iterative imputed, data,
missing positions)
mean rmse = calculate rmse(mean imputed, data, missing positions)
soft rmse = calculate rmse(soft imputed, data, missing positions)
median_rmse = calculate_rmse(median imputed, data, missing positions)
ensemble rmse = calculate rmse(ensemble imputed.values, data,
missing positions)
# Plotting the RMSEs
methods = ['KNN', 'Iterative', 'Mean', 'SoftImpute', 'Median',
'Proposed Ensemble']
```

```
rmses = [knn_rmse, iterative_rmse, mean_rmse, soft_rmse, median_rmse,
ensemble_rmse]

plt.figure(figsize=(10, 6))
plt.bar(methods, rmses, color=['blue', 'green', 'red', 'purple',
    'orange', 'cyan'])
plt.xlabel('Imputation Methods')
plt.ylabel('RMSE')
plt.title('Comparison of RMSEs for Different Imputation Methods')
plt.show()
```