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import pandas as pd
import numpy as np
from scipy.signal import savgol_filter
from sklearn.impute import KNNImputer
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import StratifiedKFold, cross_val_score
import joblib
\mbox{\tt\#} Replace 'train.csv' with your actual training file name
train = pd.read_csv('hacktrain.csv')
test = pd.read_csv('hacktest.csv')
ndvi_cols = [col for col in train.columns if '_N' in col]
def denoise_series(row):
    if row.isnull().sum() > len(row) // 2:
       return row
    filled = row.interpolate(limit_direction='both')
    window = min(7, len(row) if len(row)%2==1 else len(row)-1)
    if window < 3:
        return filled
    filtered = savgol filter(filled, window length=window, polyorder=2)
    return pd.Series(filtered, index=row.index)
X_train_denoised = train[ndvi_cols].apply(denoise_series, axis=1)
X_test_denoised = test[ndvi_cols].apply(denoise_series, axis=1)
imputer = KNNImputer(n neighbors=3)
\label{eq:colored} \textbf{X\_train\_imputed = pd.DataFrame(imputer.fit\_transform(X\_train\_denoised), columns=ndvi\_cols)}
X_test_imputed = pd.DataFrame(imputer.transform(X_test_denoised), columns=ndvi_cols)
def extract_features(row):
    ndvi = np.array(row)
    return pd.Series({
        'ndvi_mean': np.mean(ndvi),
        'ndvi_std': np.std(ndvi),
        'ndvi_min': np.min(ndvi),
        'ndvi_max': np.max(ndvi),
        'ndvi_range': np.max(ndvi) - np.min(ndvi),
        'ndvi_argmax': np.argmax(ndvi),
        'ndvi_argmin': np.argmin(ndvi),
        'ndvi_median': np.median(ndvi),
        'ndvi_q25': np.percentile(ndvi, 25),
        'ndvi_q75': np.percentile(ndvi, 75),
        'ndvi_skew': pd.Series(ndvi).skew(),
    })
X_train_features = X_train_imputed.apply(extract_features, axis=1)
X_test_features = X_test_imputed.apply(extract_features, axis=1)
# Replace 'class' with the actual name of the label column if it's different
le = LabelEncoder()
y = le.fit_transform(train['class'].astype(str))
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train_features)
X_test_scaled = scaler.transform(X_test_features)
clf = LogisticRegression(multi_class='multinomial', solver='lbfgs', max_iter=1000, class_weight='balanced', random_state=42)
cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=42)
scores = cross_val_score(clf, X_train_scaled, y, cv=cv, scoring='accuracy')
print(f'CV Accuracy: {np.mean(scores):.4f} ± {np.std(scores):.4f}')
clf.fit(X_train_scaled, y)
```

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/usr/local/lib/python3.11/dist-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning: 'multi_class' was deprecated in versi
       warnings.warn(
     CV Accuracy: 0.4785 \pm 0.0102
     /usr/local/lib/python3.11/dist-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning: 'multi_class' was deprecated in versi
       warnings.warn(
                          LogisticRegression
     LogisticRegression(class_weight='balanced', max_iter=1000,
                         multi_class='multinomial', random_state=42)
joblib.dump(clf, 'logreg_model.pkl')
joblib.dump(le, 'label_encoder.pkl')
joblib.dump(scaler, 'scaler.pkl')

    ['scaler.pkl']
y_test_pred = clf.predict(X_test_scaled)
```

⇒ submission.csv created!

print('submission.csv created!')

submission = pd.DataFrame({
 'ID': test['ID'],
 'class': y_test_labels

})

y_test_labels = le.inverse_transform(y_test_pred)

submission.to_csv('submission.csv', index=False)