```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import OneHotEncoder
from sklearn.pipeline import Pipeline
from fairlearn.metrics import (
    MetricFrame.
    selection_rate,
    demographic_parity_difference,
    equalized_odds_difference
from sklearn.metrics import accuracy score
# 1. Load dataset
df = pd.read csv("fairness dataset.csv")
# 2. Features, target, and sensitive attributes
X = df[["experience", "test_score", "interview_score", "gender"]]
y = df["hired"]
# Choose sensitive attribute for bias detection (change to "ethnicity" if present)
sensitive_feature = df["gender"]
# 3. Train/test split
X_train, X_test, y_train, y_test, sf_train, sf_test = train_test_split(
    X, y, sensitive_feature, test_size=0.3, random_state=42
# 4. Build pipeline (OneHotEncode + Logistic Regression)
pipeline = Pipeline([
    ("encoder", OneHotEncoder(drop="first")),
    ("model", LogisticRegression(solver="liblinear"))
1)
# 5. Train model
pipeline.fit(X_train, y_train)
# 6. Predictions
y_pred = pipeline.predict(X_test)
# 7. Fairness metrics by group
metric frame = MetricFrame(
    metrics={
        "accuracy": accuracy_score,
        "selection_rate": selection_rate
    },
    y_true=y_test,
    y_pred=y_pred,
    sensitive_features=sf_test
)
print("===== Overall Metrics =====")
print(metric_frame.overall)
print("\n===== Metrics by Group =====")
print(metric_frame.by_group)
# 8. Fairness difference metrics
dp_diff = demographic_parity_difference(y_test, y_pred, sensitive_features=sf_test)
eo_diff = equalized_odds_difference(y_test, y_pred, sensitive_features=sf_test)
print(f"\nDemographic Parity Difference: {dp_diff:.4f}")
print(f"Equalized Odds Difference: {eo_diff:.4f}")
    ===== Overall Metrics =====
     accuracy
                       0.500000
     selection_rate
                       0.533333
     dtype: float64
     ==== Metrics by Group =====
             accuracy selection_rate
     gender
     female
                  0.5
                             0.625000
                             0.428571
     male
                  0.5
     Demographic Parity Difference: 0.1964
     Equalized Odds Difference: 0.1964
```