!pip install fairlearn → Collecting fairlearn Downloading fairlearn-0.12.0-py3-none-any.whl.metadata (7.0 kB) Requirement already satisfied: numpy>=1.24.4 in /usr/local/lib/python3.11/dist-packages (from fairlearn) (2.0.2) Requirement already satisfied: pandas>=2.0.3 in /usr/local/lib/python3.11/dist-packages (from fairlearn) (2.2.2) Requirement already satisfied: scikit-learn>=1.2.1 in /usr/local/lib/python3.11/dist-packages (from fairlearn) (1.6.1) Requirement already satisfied: scipy>=1.9.3 in /usr/local/lib/python3.11/dist-packages (from fairlearn) (1.16.1) Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas>=2.0.3->fairlearn) (2.9.0. Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas>=2.0.3->fairlearn) (2025.2) Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas>=2.0.3->fairlearn) (2025.2) Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn>=1.2.1->fairlearn) (1.5.1) Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn>=1.2.1->fairlearn) (3. Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->pandas>=2.0.3->fairlear Downloading fairlearn-0.12.0-py3-none-any.whl (240 kB) - 240.0/240.0 kB **4.4** MB/s eta 0:00:00 Installing collected packages: fairlearn Successfully installed fairlearn-0.12.0 import pandas as pd import numpy as np 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 from fairlearn.reductions import ExponentiatedGradient, DemographicParity np.random.seed(42) experience = np.random.choice(["low", "medium", "high"], n) test score = np.random.randint(0, 11, n) interview\_score = np.random.randint(0, 11, n) gender = np.random.choice(["male", "female"], n) hired\_balanced = [np.random.choice([0, 1], p=[0.5, 0.5]) for \_ in range(n)] df\_balanced = pd.DataFrame({ "experience": experience, "test\_score": test\_score, "interview\_score": interview\_score, "gender": gender, "hired": hired\_balanced }) hired\_imbalanced = [] for g in gender: if g == "male": hired\_imbalanced.append(np.random.choice([0, 1], p=[0.3, 0.7])) else: hired\_imbalanced.append(np.random.choice([0, 1], p=[0.7, 0.3])) df\_imbalanced = pd.DataFrame({ "experience": experience, "test score": test score, "interview\_score": interview\_score, "gender": gender, "hired": hired\_imbalanced }) def evaluate\_fairness(df, title, model): X = df[["experience", "test\_score", "interview\_score", "gender"]] y = df["hired"]

sensitive\_feature = df["gender"]

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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
   model.fit(X_train, y_train)
   y_pred = model.predict(X_test)
   metric_frame = MetricFrame(
        metrics={"accuracy": accuracy_score, "selection_rate": selection_rate},
       y_true=y_test,
       y_pred=y_pred,
        sensitive_features=sf_test
   \label{eq:def-def-def} $$ dp\_diff = demographic\_parity\_difference(y\_test, y\_pred, sensitive\_features=sf\_test) $$
   eo_diff = equalized_odds_difference(y_test, y_pred, sensitive_features=sf_test)
   print(f"\n===== {title} Dataset =====")
   print("Overall Metrics:", metric_frame.overall)
   print("By Group Metrics:\n", metric_frame.by_group)
   print(f"Demographic Parity Difference: {dp_diff:.4f}")
   print(f"Equalized Odds Difference: {eo_diff:.4f}")
pipeline = Pipeline([
    ("encoder", OneHotEncoder(drop="first")),
    ("model", LogisticRegression(solver="liblinear"))
])
evaluate_fairness(df_balanced, "Balanced", pipeline)
evaluate_fairness(df_imbalanced, "Imbalanced", pipeline)
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     ==== Balanced Dataset =====
     Overall Metrics: accuracy
                                         0.483333
     selection_rate 0.416667
     dtype: float64
     By Group Metrics:
              accuracy selection_rate
     gender
     female 0.481481
                             0.296296
            0.484848
                             0.515152
     Demographic Parity Difference: 0.2189
     Equalized Odds Difference: 0.2500
     ==== Imbalanced Dataset =====
     Overall Metrics: accuracy
                                         0.700000
                      0.633333
     selection_rate
     dtype: float64
     By Group Metrics:
              accuracy selection_rate
     gender
     female 0.518519
                             0.222222
            0.848485
                             0.969697
     male
     Demographic Parity Difference: 0.7475
     Equalized Odds Difference: 0.9655
```