



```
# =====
# 1. Libraries
# =====
import numpy as np
import pandas as pd

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report

# =====
# 2. Synthetic HR Dataset
# =====
np.random.seed(42)
n = 2000

df = pd.DataFrame({
    "age": np.random.randint(21, 60, n),
    "experience_years": np.random.randint(0, 20, n),
    "attendance_rate": np.clip(np.random.normal(0.9, 0.1), 0.0, 1.0),
    "training_score": np.clip(np.random.normal(70, 10), 0.0, 100.0),
    "engagement_score": np.clip(np.random.normal(75, 15), 0.0, 100.0),
    "num_warnings_issued": np.random.choice([0,1,2,3], n),
    "department": np.random.choice([
        "Operations", "HR", "Finance", "Tech", "Support"
    ])
})

# =====
# 3. Performance Label Creation
# =====
latent_score = (
    0.3 * df["attendance_rate"] * 100 +
    0.25 * df["training_score"] +
    0.2 * df["engagement_score"] -
    5 * df["num_warnings_issued"]
)
```



```
)  
  
df["performance_label"] = pd.qcut(  
    latent_score,  
    q=[0, 0.4, 0.75, 1.0],  
    labels=["Low", "Average", "High"]  
)  
  
# =====  
# 4. Preprocessing  
# =====  
dept_encoder = LabelEncoder()  
df["department"] = dept_encoder.fit_transform(df[ "departme...  
  
X = df.drop("performance_label", axis=1)  
y = df["performance_label"]  
  
# =====  
# 5. Train-Test Split  
# =====  
X_train, X_test, y_train, y_test = train_test_split(  
    X, y, test_size=0.2, random_state=42  
)  
  
# =====  
# 6. Model Training (ML Core)  
# =====  
rf_model = RandomForestClassifier(  
    n_estimators=100,  
    random_state=42  
)  
  
rf_model.fit(X_train, y_train)  
  
# =====  
# 7. Model Evaluation  
# =====  
y_pred = rf_model.predict(X_test)
```



```
print("Accuracy:", accuracy_score(y_test, y_pred))
print("\nClassification Report:\n")
print(classification_report(y_test, y_pred))
print("\nConfusion Matrix:\n")
print(confusion_matrix(y_test, y_pred))

# =====
# 8. Sample Case Prediction (FINAL DEMO PART)
# =====
sample_employee = pd.DataFrame({
    "age": [34],
    "experience_years": [7],
    "attendance_rate": [0.94],
    "training_score": [78],
    "engagement_score": [82],
    "num_warnings_issued": [0],
    "department": dept_encoder.transform(["Operations"])
})

sample_prediction = rf_model.predict(sample_employee)

print("\nPredicted Performance for Sample Employee:", :)
```

Accuracy: 0.92

Classification Report:

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| Average | 0.85 | 0.94 | 0.89 | 141 |
| High | 0.96 | 0.91 | 0.93 | 93 |
| Low | 0.97 | 0.91 | 0.94 | 166 |
| accuracy | | | 0.92 | 400 |
| macro avg | 0.92 | 0.92 | 0.92 | 400 |
| weighted avg | 0.92 | 0.92 | 0.92 | 400 |

Confusion Matrix:

```
[[132  4  5]
 [ 8  85  0]
 [15   0 151]]
```



Predicted Performance for Sample Employee: High

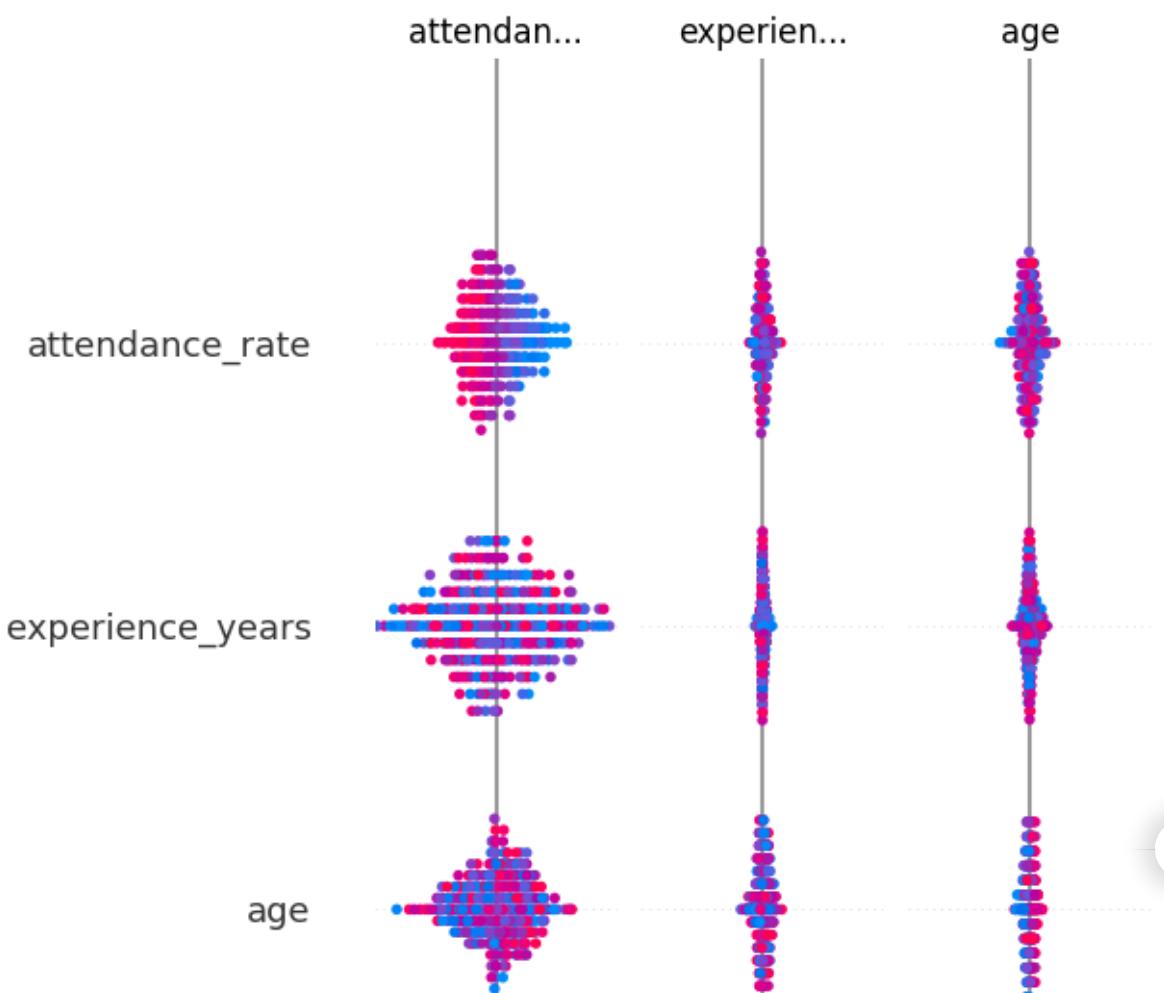
```
import shap

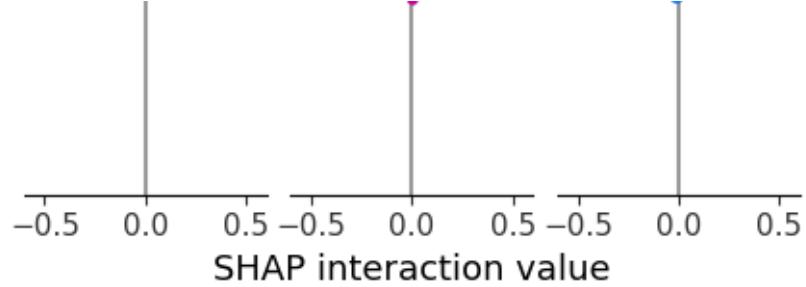
# Create SHAP explainer
explainer = shap.TreeExplainer(model)

# Calculate SHAP values
shap_values = explainer.shap_values(X_test)

# Global feature importance plot
shap.summary_plot(shap_values, X_test)
```

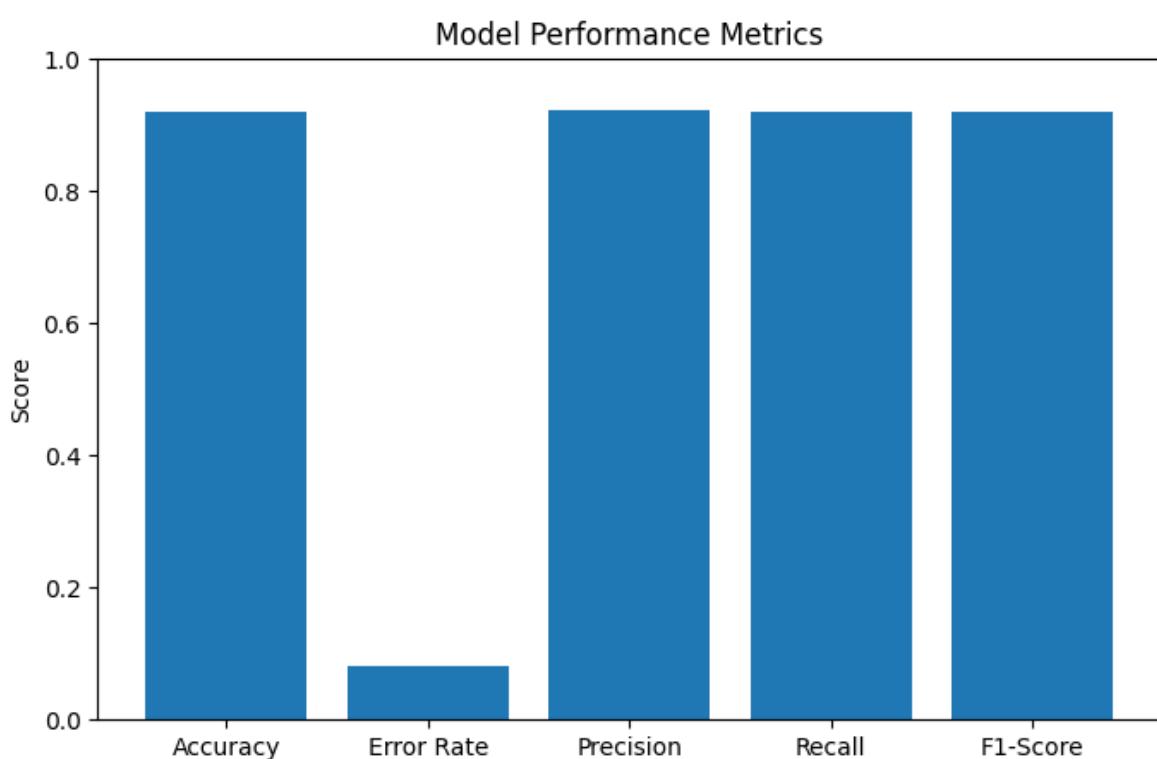
```
/tmp/ipython-input-3036699843.py:10: FutureWarning: The NumPy global RNG was
  shap.summary_plot(shap_values, X_test)
/usr/local/lib/python3.12/dist-packages/shap/plots/_beeswarm.py:723: FutureWa
  summary_legacy(
/usr/local/lib/python3.12/dist-packages/shap/plots/_beeswarm.py:743: FutureWa
  summary_legacy(
/usr/local/lib/python3.12/dist-packages/shap/plots/_beeswarm.py:743: FutureWa
  summary_legacy(
```





```
import matplotlib.pyplot as plt

plt.figure(figsize=(8, 5))
plt.bar(metrics_df["Metric"], metrics_df["Value"])
plt.title("Model Performance Metrics")
plt.ylabel("Score")
plt.ylim(0, 1)
plt.show()
```



Start coding or generate with AI.

