



EMPLOYEE PERFORMANCE PREDICTION :USING MACHINE LEARNING



PROGRESS REPORT

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WHY I CHOSE THIS PROBLEM

Employee performance is one of the most critical factors determining an organization's success. In today's competitive business environment, organizations are constantly looking for ways to retain top talent, improve productivity, and make data-driven HR decisions. However, traditional performance evaluations are often

Subjective

Time-consuming

With my background in law and business, This problem combines HR strategy with advanced technologies like machine learning, offering a meaningful solution to improve workforce planning and talent management. It can also support equity and transparency in performance evaluations, which is a major challenge in modern workplaces.

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| Author Name | Year of Publication | Ti tle | Method ology | Accuracy | Research Gaps |
|-----------------------------------|---------------------|--|--|------------------------------|---|
| Pablo Robles-Granda et al. | 2023 | Jointly Predicting Job Performance, Personality, Cognitive Ability, Affect, and Well-Being | Behavioraltracking, mobile and wearable data, psychological surveys | Not explicitly mentioned | Limited to U.S.-based knowledge workers; may not generalize globally |
| MD Rokibul Hasan et al. | 2022 | Employee Performance Prediction: An Integrated Approach of Business Analytics and Machine Learning | Business analytics, machine learning (feature engineering, model tuning) | High (exact % not stated) | Focuses on structured data; ignores unstructured data like feedback or commu ni cati on |
| Dr. Kishan Singh & Ayesha Khuteja | 2021 | Employee Performance Prediction Using Machine Learning Algorithms | CRISP-DM framework, Random Forest classification | 87% | Limited algorithm diversity; does not explore ensemble or deep learning |
| –(Conference Paper) | 2022 | Application of Machine Learning in HRM: Employee Performance Prediction Model | ML models (Linear Regression, SVM, Decision Tree, Random Forest) | Random Forest performed best | Based on a single company’s data; lacks model interpretability and transparency |

FORNT-END PLANS

Dashboard: Overview of employee performance predictions.

Employee Profile: Detailed view with performance trends.

Upload Feature: Upload employee data (CSV/XLSX).

Search & Filter: Find employees by role, department, or performance level.

Reports: Option to download individual performance reports (PDF).



```
[5]: import pandas as pd
from sklearn.ensemble import RandomForestRegressor
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error
import gradio as gr

# Load dataset
df = pd.read_csv("train_dataset.csv")

# Handle missing values
df.dropna(inplace=True)

# Separate features and target
X = df.drop("actual_productivity", axis=1)
y = df["actual_productivity"]

# Train/test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Model training
model = RandomForestRegressor(n_estimators=100, random_state=42)
model.fit(X_train, y_train)

# Evaluation
mse = mean_squared_error(y_test, model.predict(X_test))

# Prediction function
def predict_productivity(*features):
    input_df = pd.DataFrame([features], columns=X.columns)
    prediction = model.predict(input_df)[0]
    return round(prediction, 4)

# Create Gradio inputs dynamically
inputs = []
for col in X.columns:
    if X[col].nunique() < 10:
        input_type = "text" if X[col].dtype == "object" else "number"
        inputs.append(
            gr.inputs.Textbox if input_type == "text" else gr.inputs.Number(
                min=X[col].min(), max=X[col].max(), step=1, value=X[col].mean()
            )
        )
    else:
        inputs.append(gr.inputs.Choice(choices=X[col].unique(), value=X[col].mode()[0]))

output = gr.outputs.Number()

gr.Interface(
    predict_productivity,
    inputs,
    output,
    title="Productivity Predictor",
    description="A tool to predict productivity based on various features."
).launch()
```

Conclusion - this project demonstrates the use of machine learning to predict employee productivity based on historical data using a random forest regressor. This project highlights how companies can transform raw employee data into actionable insights that improve productivity, efficiency and decision making.