

Adult Income Analysis

Predicting Income Levels Based on Personal Features

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Understanding Our Stakeholder

Our stakeholder is the marketing department for a firm and aims to understand the income level of adults based on their personal features

This analysis will assist in designing effective marketing strategies



About the Data

The data is comprised of 48,842 adults with features including education, marital status, race, and gender

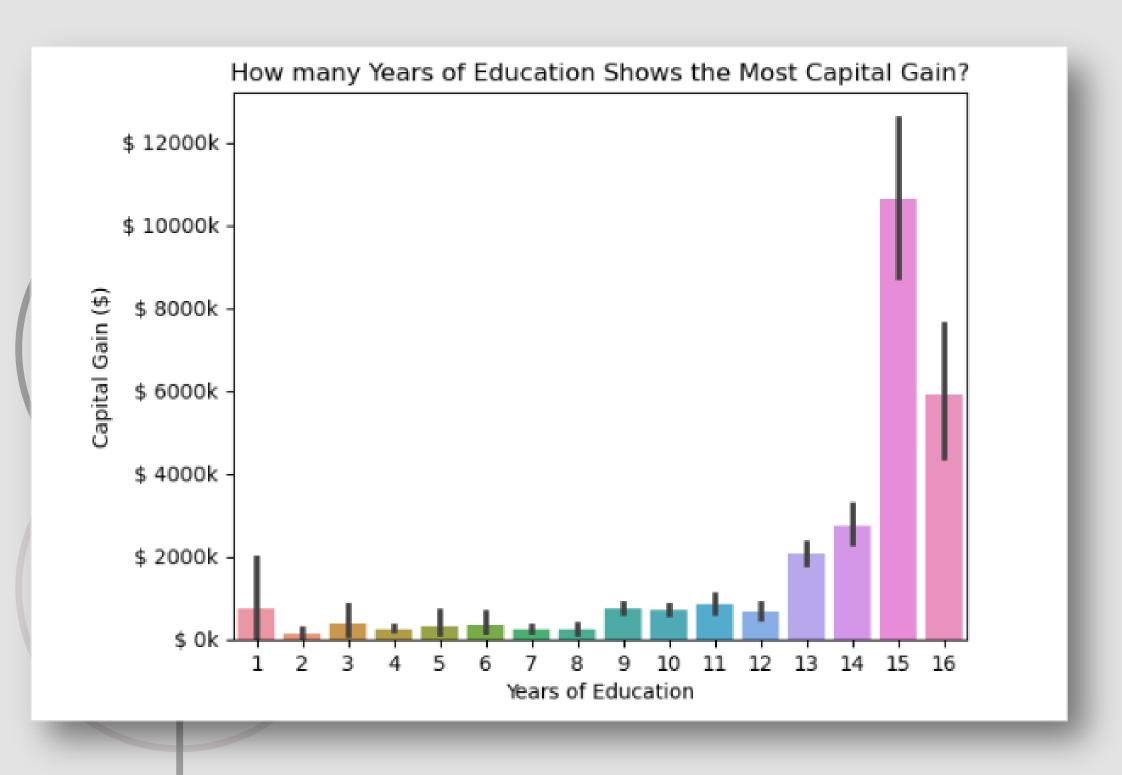
The target variable is income, categorized as >50k and <=50k



Key Finding #1: Education and Capital Gain

A correlation exists between years of education and capital gain

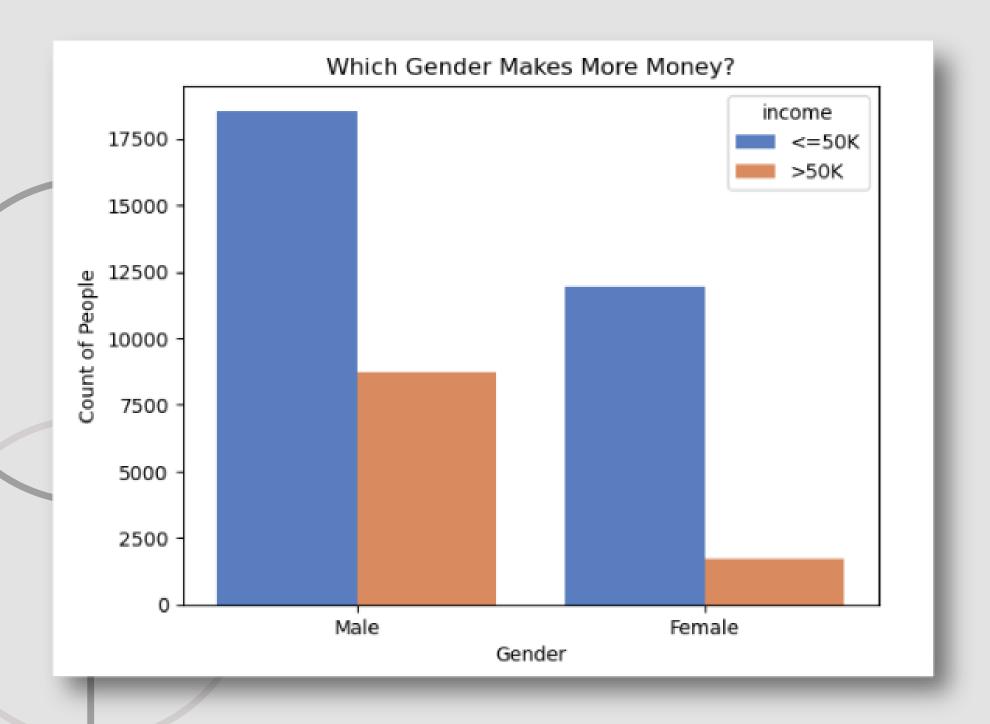
Increased education levels tend to lead to higher capital gain, dropping off at 16 years of education



Key Finding #2: Gender and Income Level

Gender also correlates with income level, with males typically earning more than females

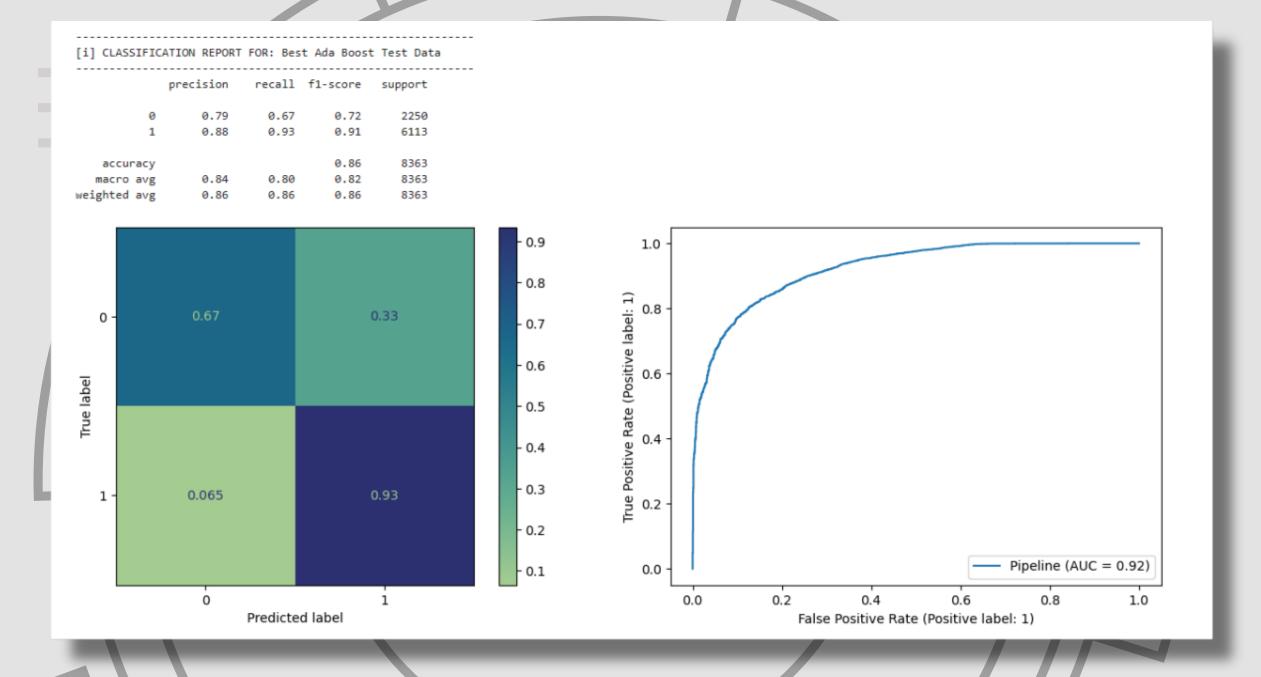
The disparity in income distribution between genders is significant



Our Predictive Model: Ada Boosted Model

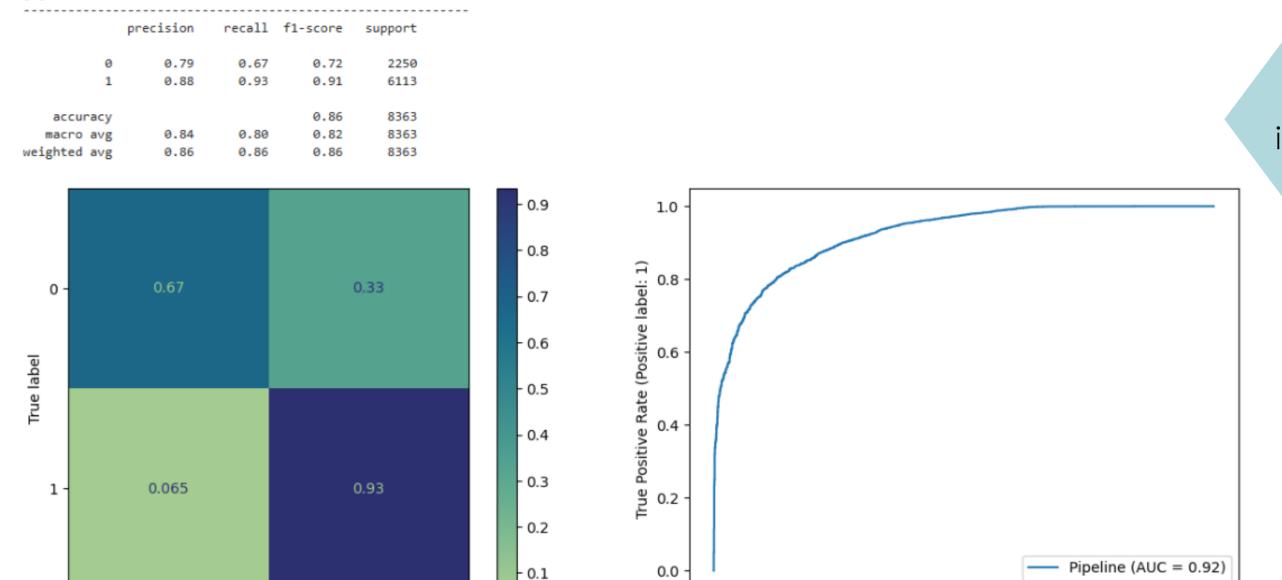
Our chosen predictive model is an Ada Boosted Model with Tuned Hyperparameters

Compared to other models such as Random Forest and KNN, this model yielded the highest metrics



Understanding Our Model's Performance

The model is 86% accurate, performing best for predicting incomes <=50k



Predicted label

0.0

0.2

0.4

False Positive Rate (Positive label: 1)

0.8

1.0

The model has a 93% success rate for predicting <=50k incomes and a 67% success rate for predicting >50k incomes

Recommendations and Limitations

Recommendation

Use this model for real-world analysis, especially for adults earning under 50k



Limitation

The model's data is primarily from the US and consists predominantly of white individuals

