

# Who is my target audience?

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# Overview

## **Stakeholder:**

Amy White, Marketing Manager

## **Business Problem:**

Amy is tasked with identifying keywords and demographic information to use as part of a paid marketing push for an online class to improve your career.

She wants to target individuals making less than \$50K a year.

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# Selecting a Data Set

## US Census Data

- For this project, I chose a data set from Kaggle that came from the US Census. This data set has the following set of features:
- The original data set had 32,561 entries and 15 columns. The data set used here had 32,537 entries and 13 columns after dropping irrelevant information.

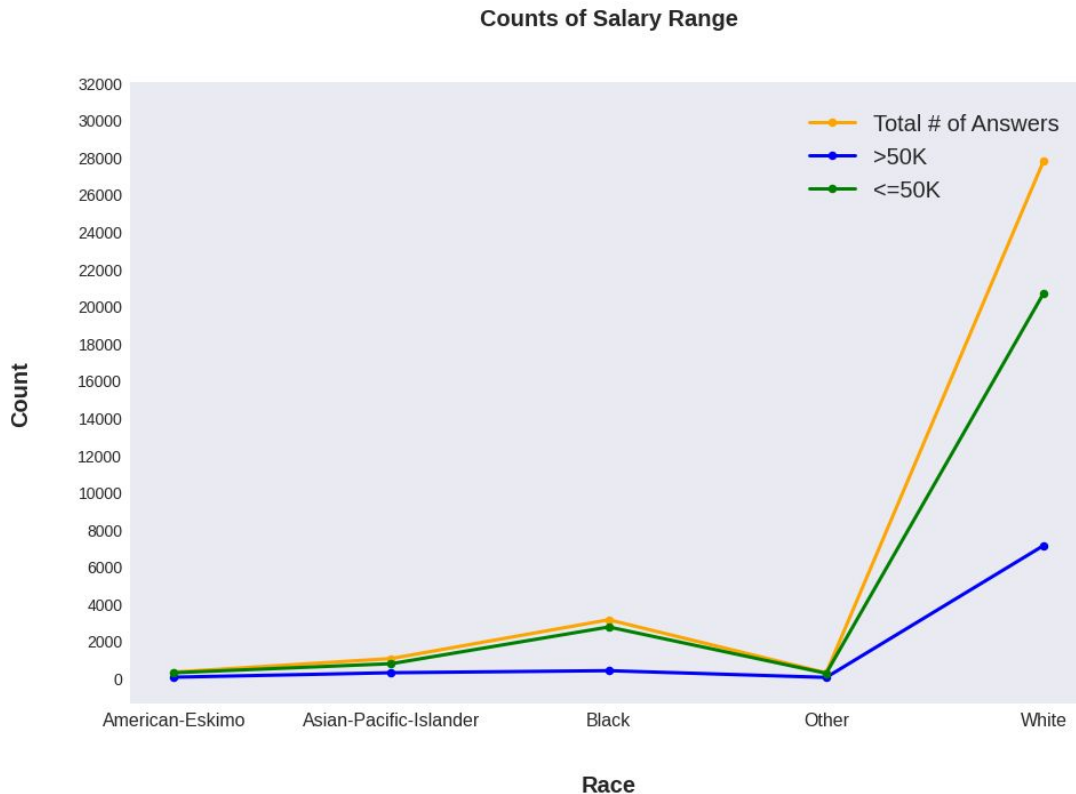
#	Column	Non-Null Count	Dtype
0	age	32537 non-null	int64
1	workclass	32537 non-null	object
2	education	32537 non-null	object
3	marital_status	32537 non-null	object
4	occupation	32537 non-null	object
5	relationship	32537 non-null	object
6	race	32537 non-null	object
7	gender	32537 non-null	object
8	capital_gain	32537 non-null	int64
9	capital_loss	32537 non-null	int64
10	hours_per_week	32537 non-null	int64
11	native_country	32537 non-null	object
12	outcome	32537 non-null	object

# Visual 1 - Line Graph of Salaries by Race

This line plot represents salaries by race. This census data categorized salaries as either >50K or <=50K.

We can see there is a significant gap between the orange line and the blue line in this category, indicating that the overwhelming majority of black people represented here make less than \$50K.

Ultimately what this tells us, is that race is not a strong indicator of salary and shouldn't be used as a keyword.

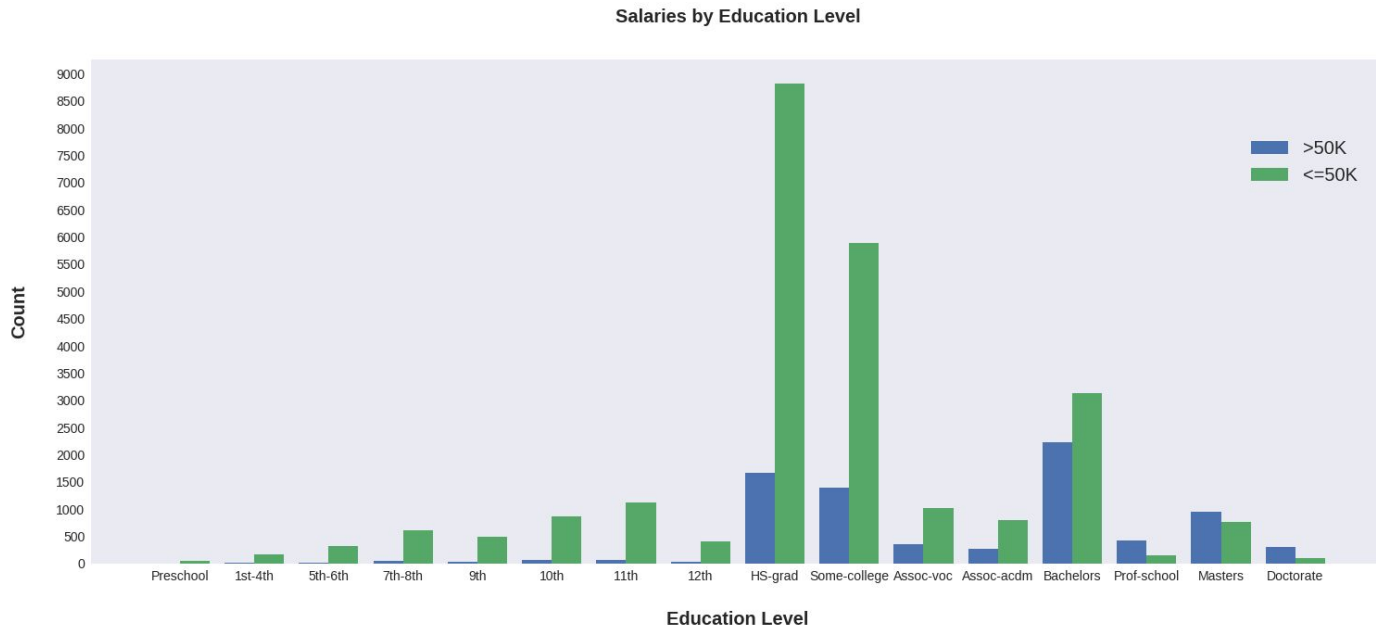


# Visual 2 - Bar Chart of Salaries by Education

This bar chart represents salaries by education level. As we can clearly see it pays to stay in school! The highest concentration of salaries >\$50K is from HS grad and beyond.

Prof school, masters, and doctorate, are the only columns where those earning >\$50k outnumber those making <=\$50K.

Meaning Bachelors degree and under will make the best keywords.



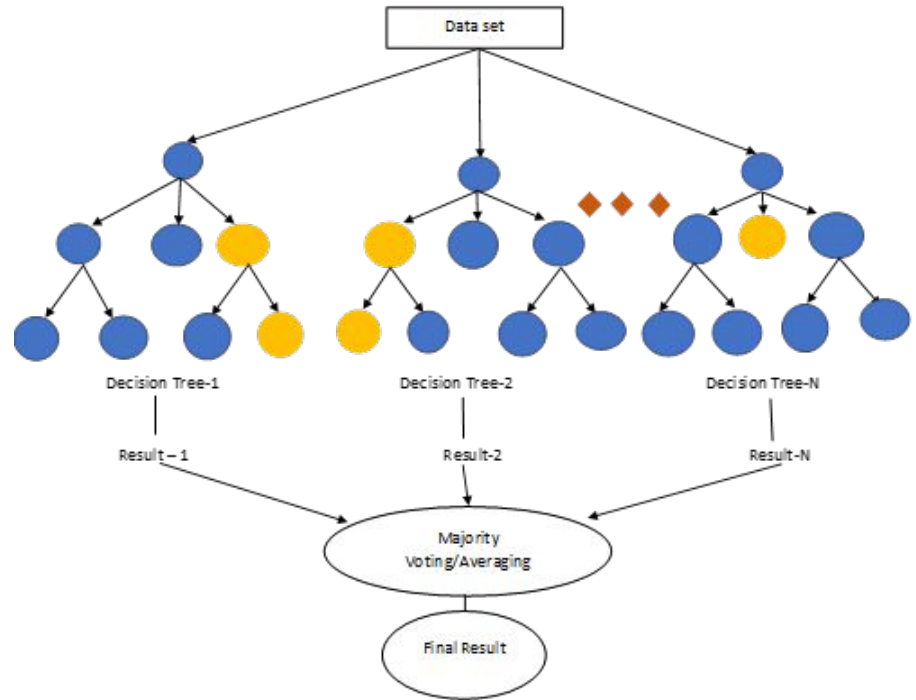
This visual gives us a much better idea of what keywords would be useful.

# Machine Learning

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# Model - Random Forest Classifier

Using the insights gathered from the explanatory visuals, it was decided to build a Random Forest Classifier algorithm to find patterns throughout the data and predict whether or not someone makes  $\leq \$50K$  based off of demographic features.

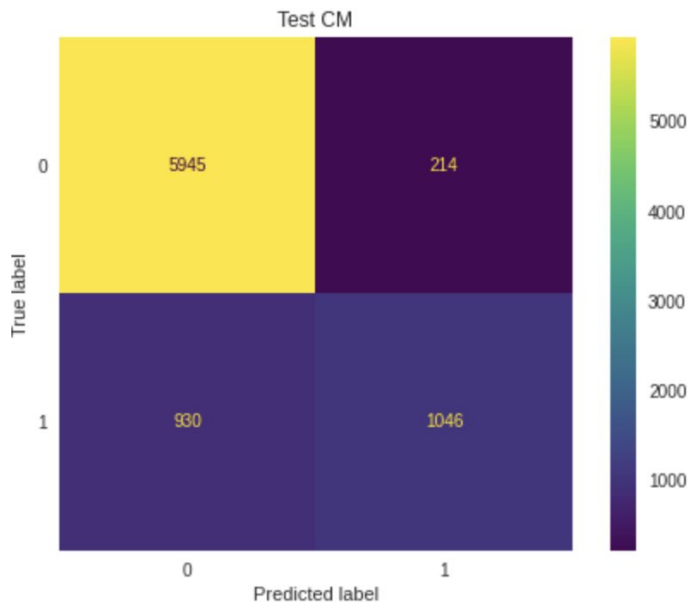


# Performance Metrics

One strength of this model is that it scores highly in precision and accuracy on the  $\leq \$50K$  outcome set. Since this is the audience we are targeting, the performance on the  $> \$50K$  outcome set isn't as important.

It was with this model we were able to achieve our highest number of True Negatives. A true negative in this case means that we predicted that they would make less than \$50K, and in fact they do.

In summary, our model is 86% accurate when it comes to predicting if someone makes less than \$50K.





# Summary

1. This model can successfully predict whether someone makes less than \$50K based off of demographic information.
  2. Using a combination of education, workclass, and occupation features as keywords will result in the target audience.
  3. Be cautious when using demographic data to ensure results are equitable.
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