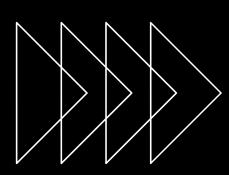
Discovering

who the 2 Group is

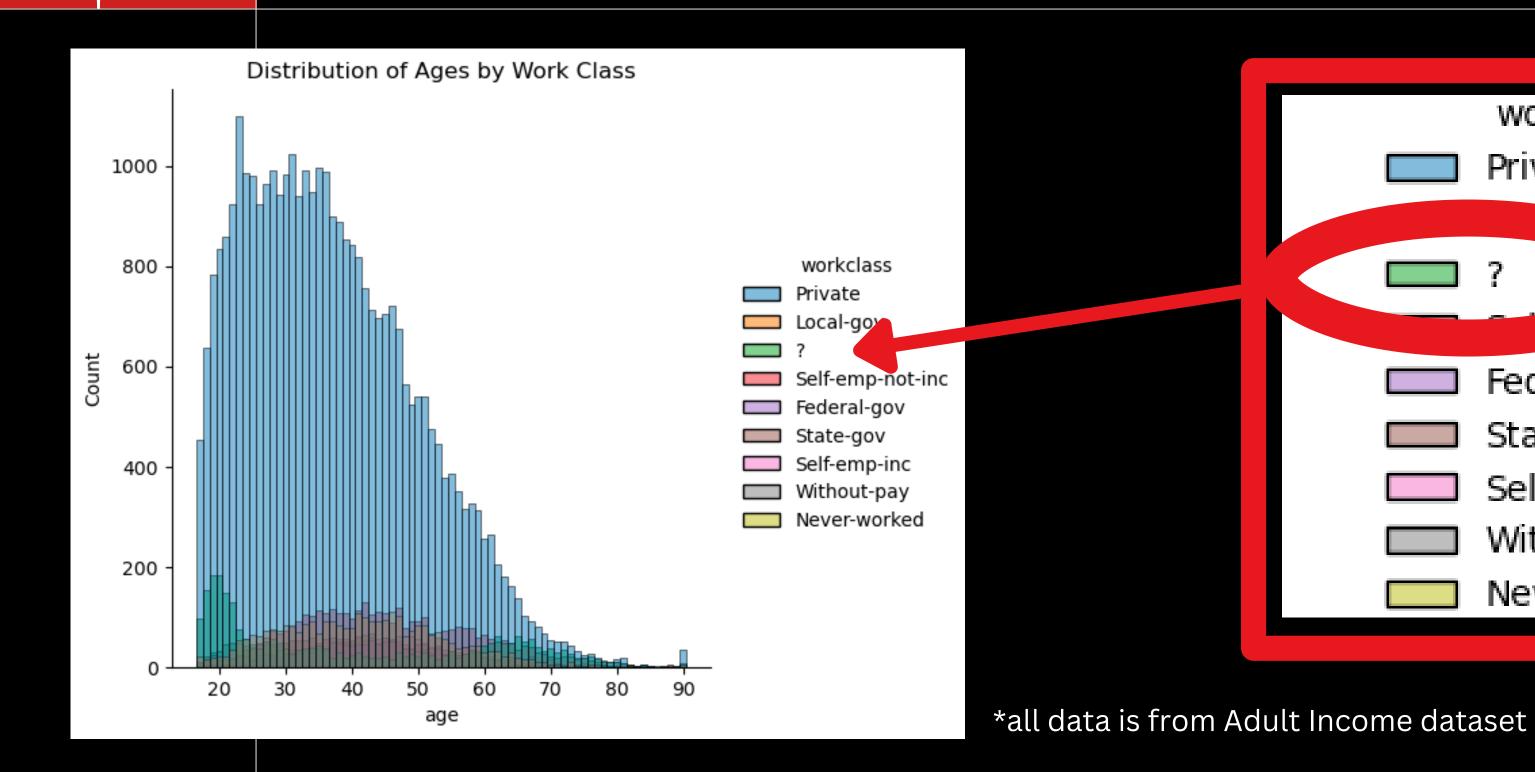
An experiential movement through and about one demographic

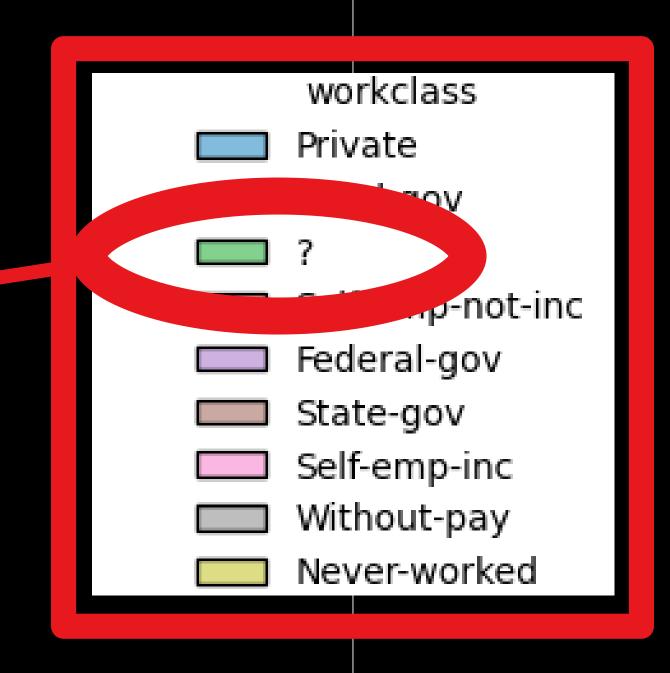




WHAT IS THE ? GROUP

From a cursory look at age demographics broken up by work sector, there is a curious category called?





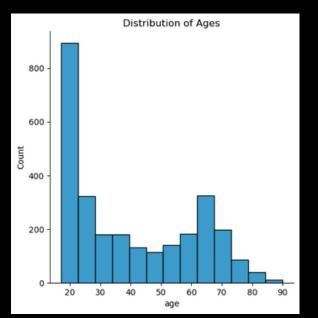
The ? Group

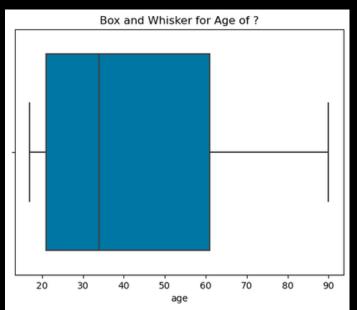
This group is left-skewed in relation to age distribution. Although the average age of the ? Group is 40.1 years (compared to the sample's 38.6), they also have a median age of 34 years (below that of the samples 37 years)

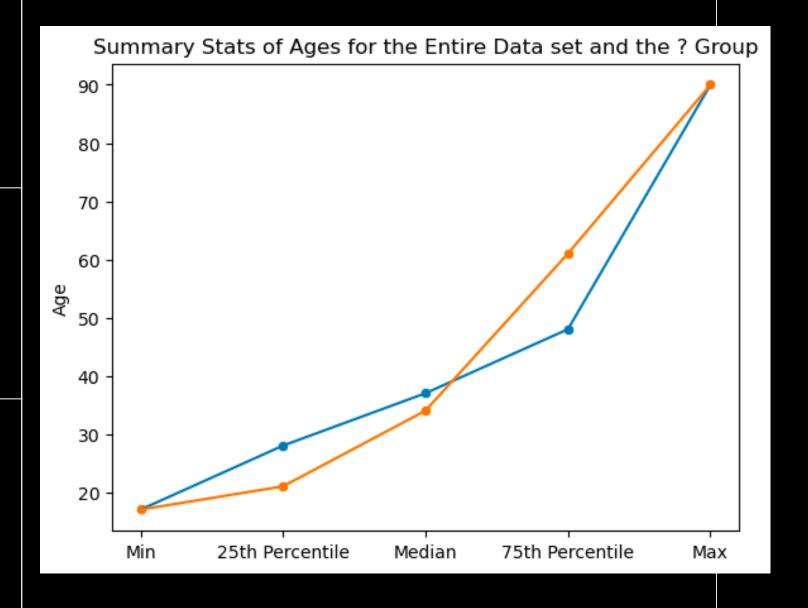
The ? Group is as 1/20th the size of the entire dataset. It closely mirrors the racial demographics of the whole. <u>82.6% of ? are White</u> (compared to 85.5% (-2.9%)), <u>11.6% are Black</u> (9.6% (+2%)), <u>3.4% are Asian/Pacific-Islander</u> (3.1% (+0.3%)), <u>1.3% are American-Indian/Eskimo</u> (1.0% (+0.3%)), and <u>1.1% are Other</u> (0.8% (+0.3%)).

The gender differences are quite large. The ? Group is close to evenly split between binary genders: 54.6% male, 45.4% female. The overall dataset is 66.8% male and 33.2% female. This is the closest split for any other work class. The second closest is the local government (59.9%M/40.1%F) and the group with the biggest split is self employed (87.6%M/12.4%F).

For the entire data set, 76.1% of the people make \$50,000 or less. For the ? Group, 90.53% make that \$50K or less. The second closest split is the without-pay group (90.47%), while the never-worked group logically are all within that income category. The group with the highest percentage of workers making more than \$50K is self-employeed, with 44.7%.







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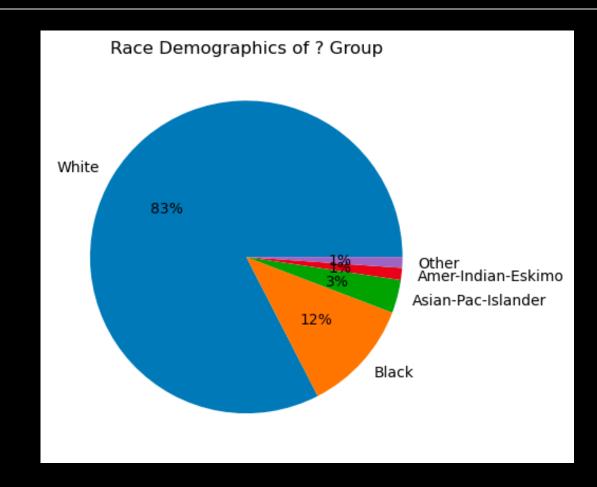
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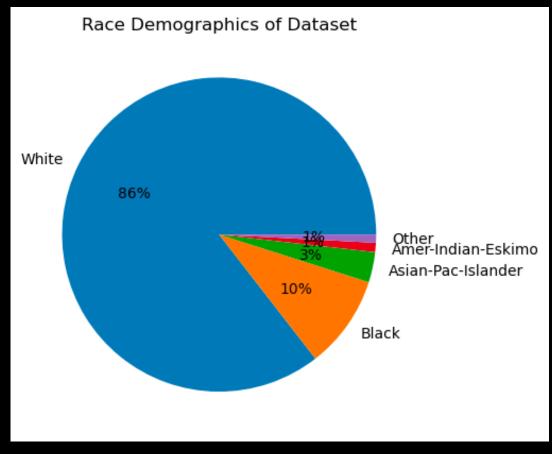
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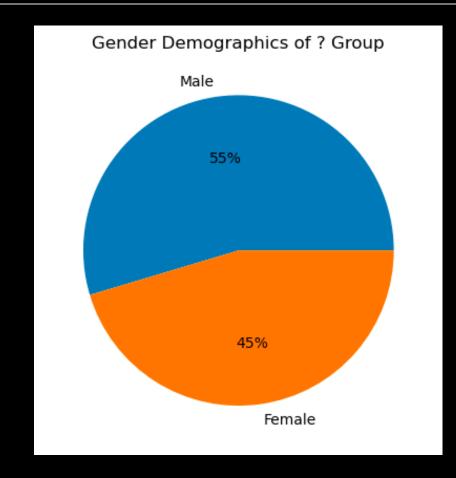
The ? Group

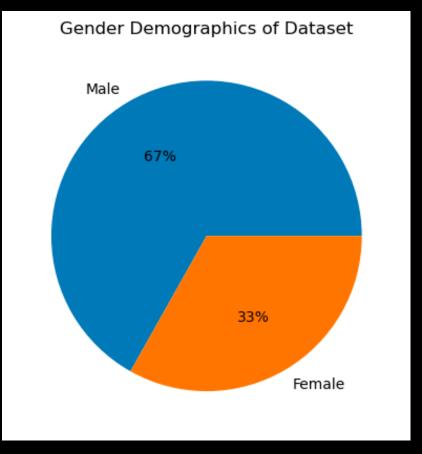
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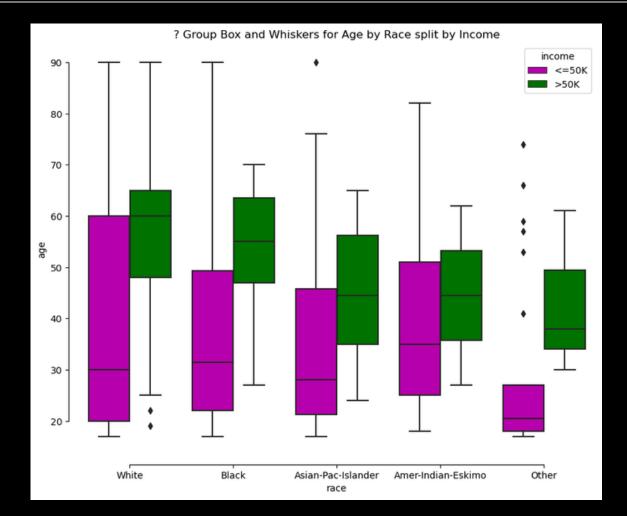
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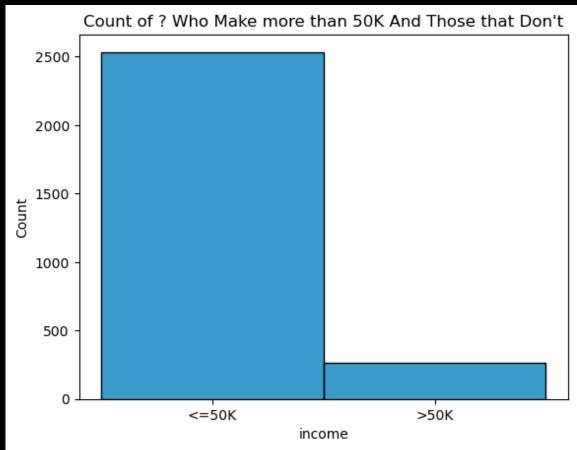
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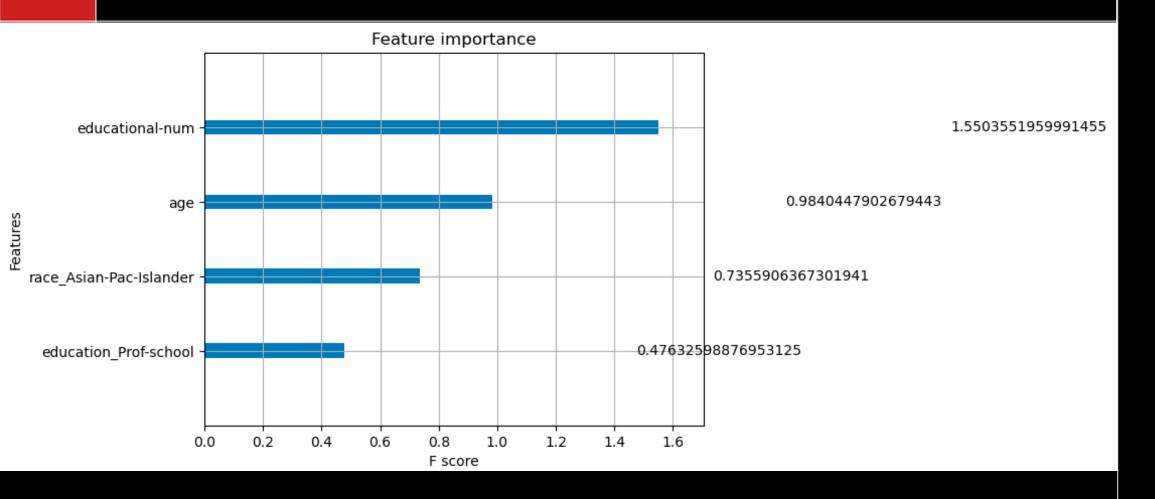
Where is the ? Group from



United-States 2530 Mexico 48 ? 46 Canada 19 South 14 Germany 13 Philippines 12 Taiwan 10 China 9 Puerto-Rico 9 El-Salvador 8 England 8 Poland 6 Dominican-Republic 6 Haiti 6 Cuba 5 Portugal 5 Italy 5 India 4 Jamaica 3 Ігап З Columbia 3 Vietnam 3 Japan 3 Guatemala

France 2
Cambodia 2
Hong 2
Laos 2
Ecuador 2
Honduras 1
Nicaragua 1
Trinadad&Tobago 1
Hungary 1
Ireland 1
Outlying-US(Guam-USVI-etc) 1
Peru 1
Thailand 1
Scotland 1

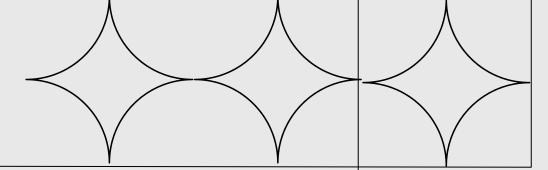
After running a machine learning classifier algorithm (XGBoost), it appears that the most importnt features are the number of educational years, their age, whether they identify their race as Asian/Pacific Island, and their school as professional



What's important for identifying someone from the? Group



Who are ? in relation to the data set



They are not making the most money, that's for sure.

