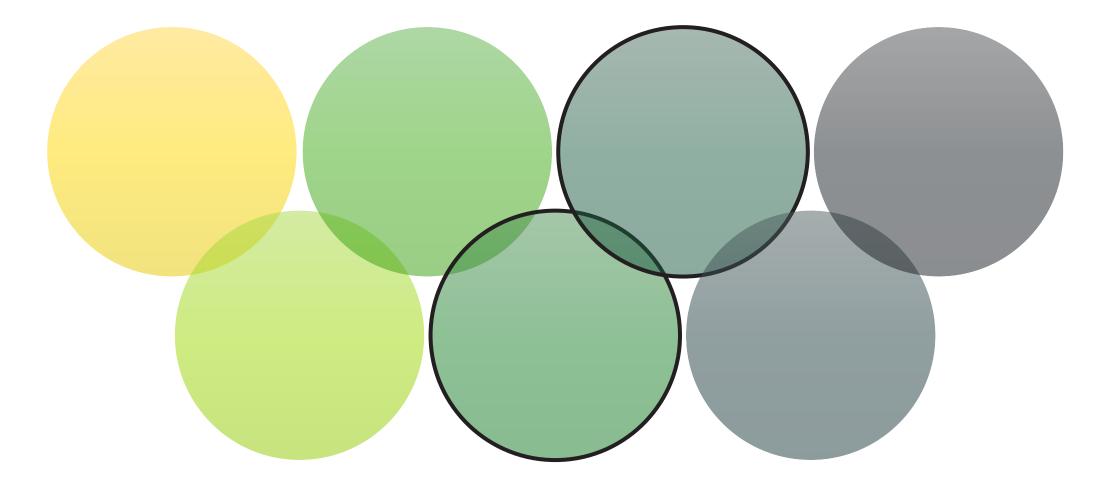
AI IN EARTH OBSERVATION

GUIDELINES & FRAMEWORKS

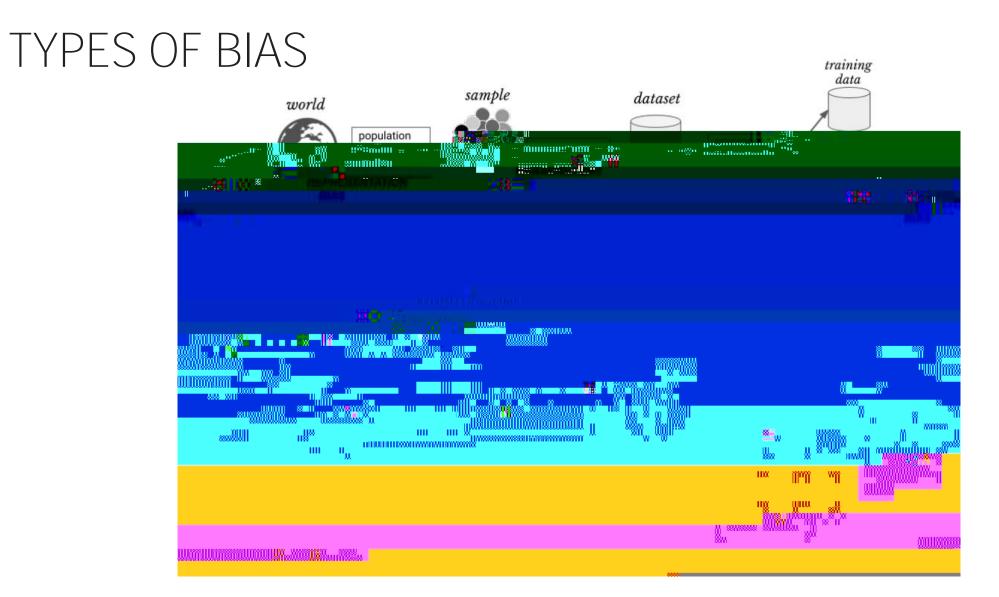




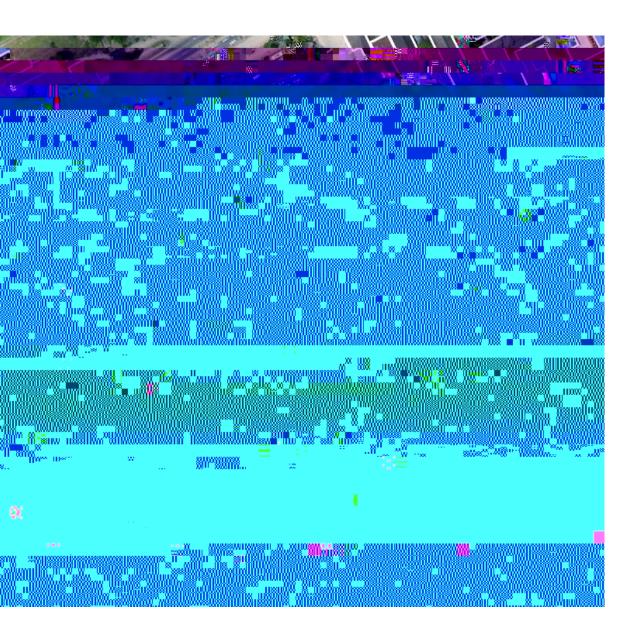
FAIRNESS / BIAS

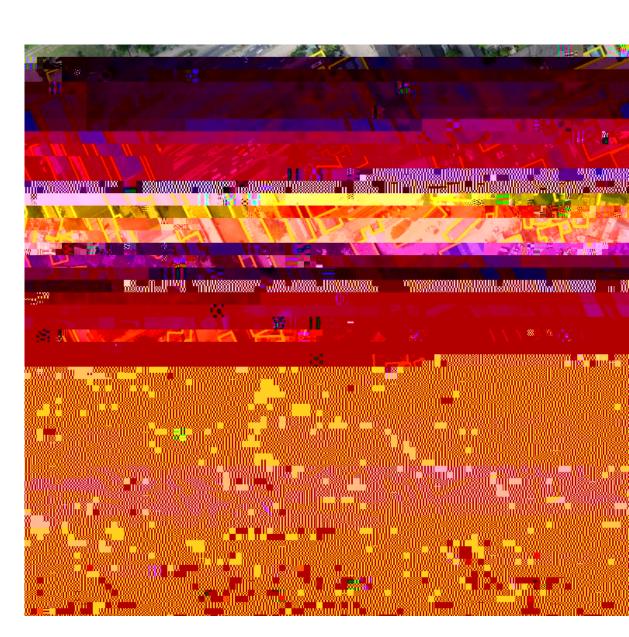
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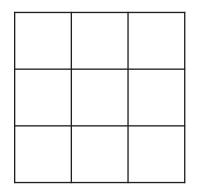


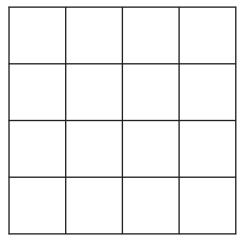


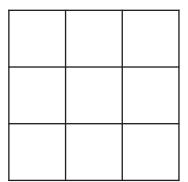


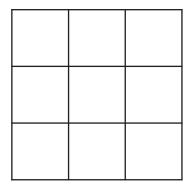
EARSeL Cyprus 2022.

FAIRNESS?









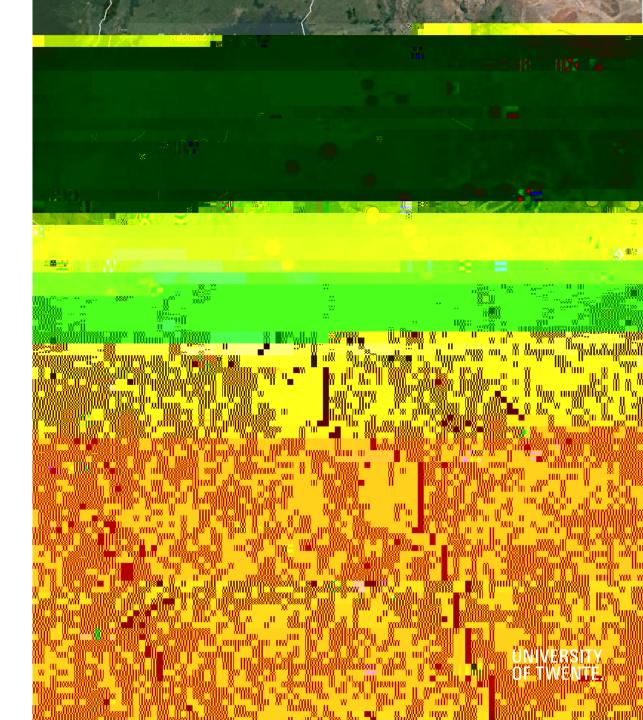
SIMPLE

MULTI-CLASS

AUDIT FOR BIASES



EXAMPLE Ì BUILDING DETECTION IN TANZANIA



EARSeL Cyprus 2022.

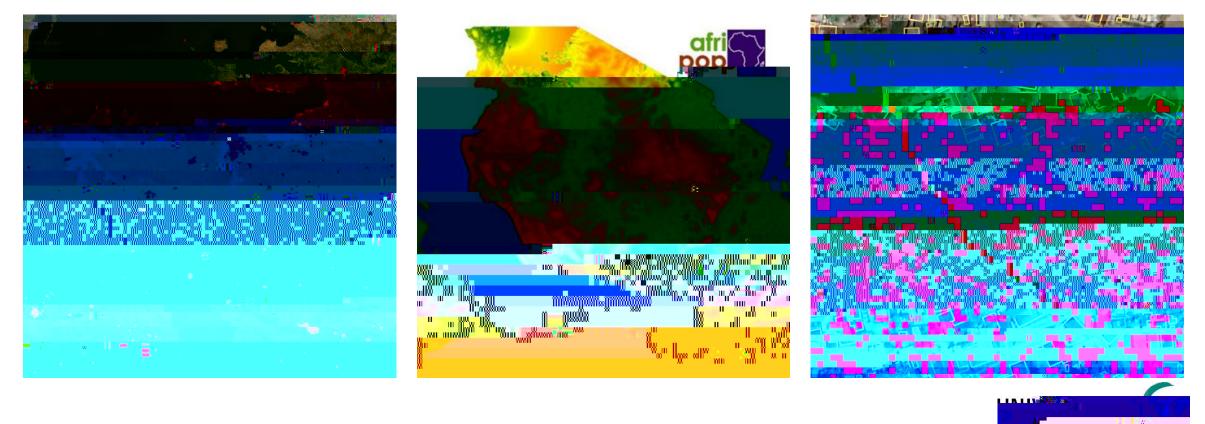
SENSITIVE ATTRIBUTES

CITY SIZE

POVERTY

BUILDING SIZE

X VLIVIL. IT



ACCURACY Ì

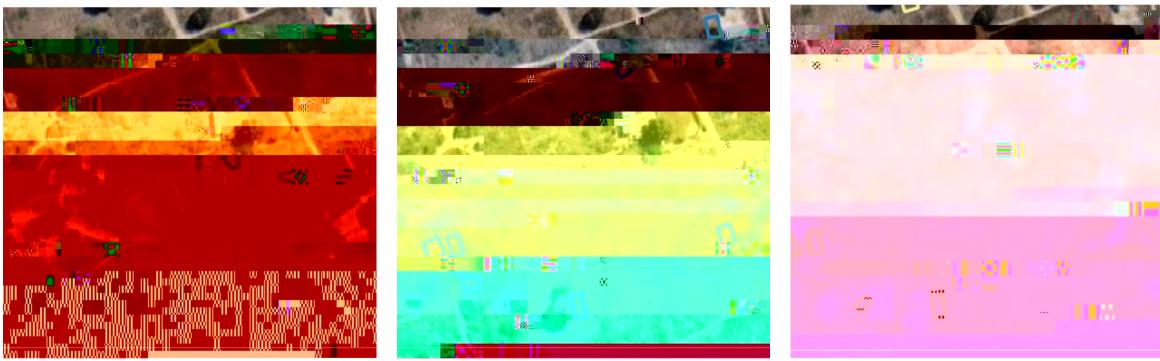


ACCURACY È EXAMPLE HIGH POVERTY, RURAL

OSM

BING

GOOGLE

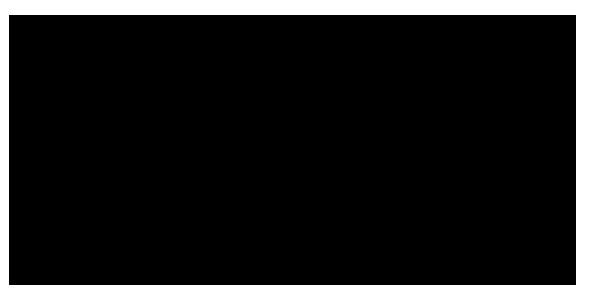




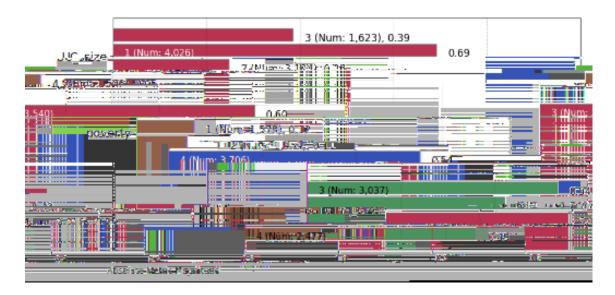
EARSeL Cyprus 2022.

PRECISION





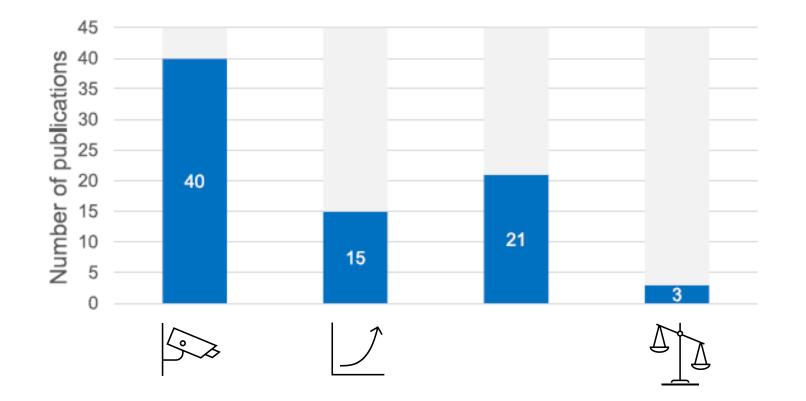
GOOGLE



EXPLAINABILITY

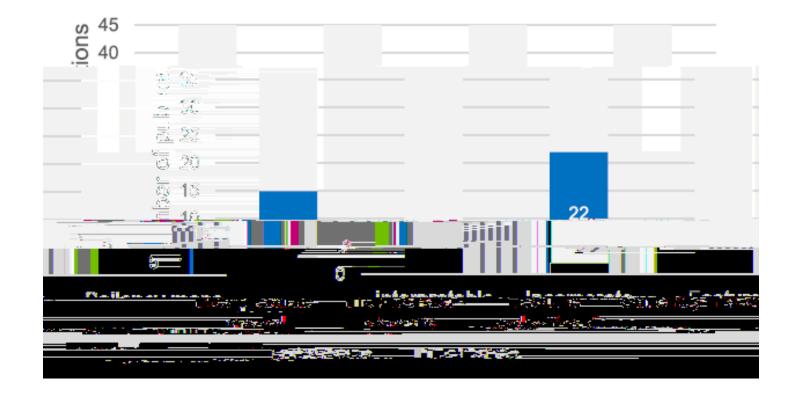


MOTIVATIONS FOR EXPLAINABLE AI





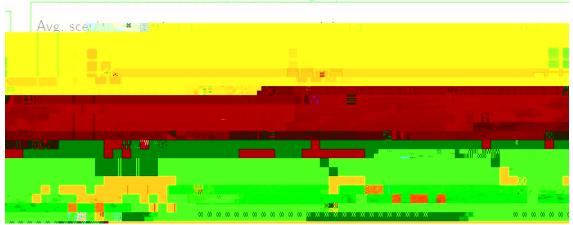
TYPES OF EXPLAINABLE AI IN EARTH OBSERVATION





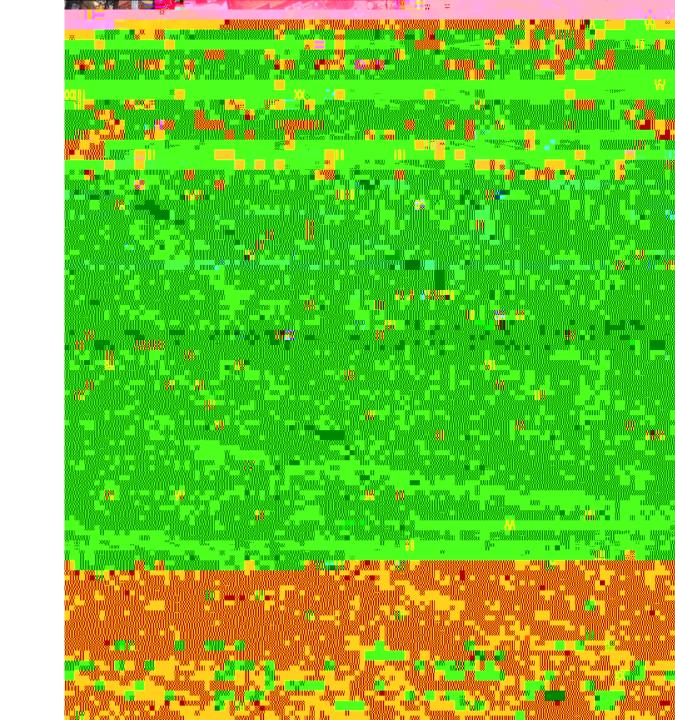
INTERPRETABLE MODELS

ScenicOrNot dataset (200k images)

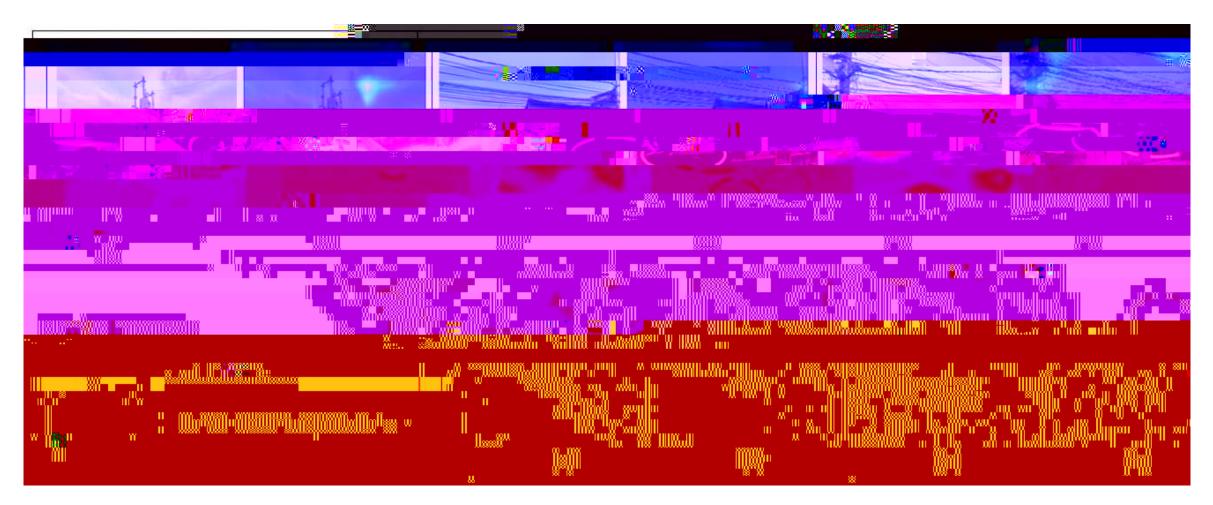




Features:



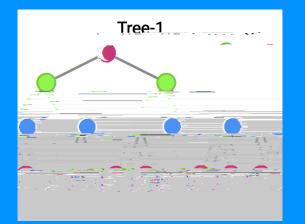
SALIENCY MAPS



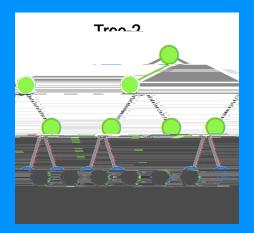


Integrating Remote Sensing and Street View Images to Map Slums Using Deep Learning Approach

ARE RANDOM FORESTS INTERPRETABLE?



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Int. Journal of Applied EO and Geoinfo

EXPLAINABILITY È IS THIS

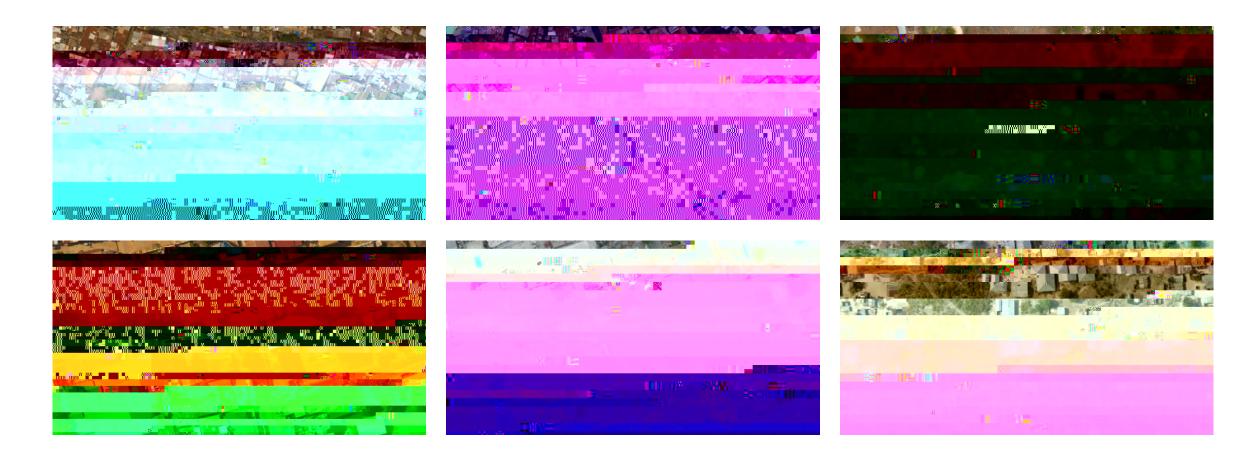
. .

GENERALIZATION CAPABILITY

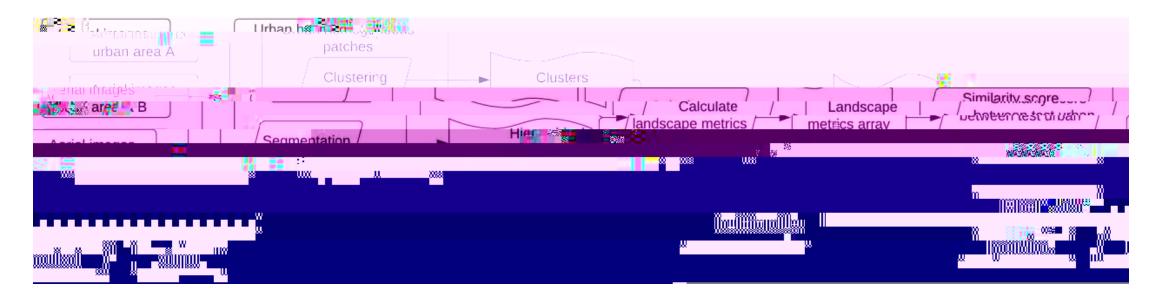
ACCRA

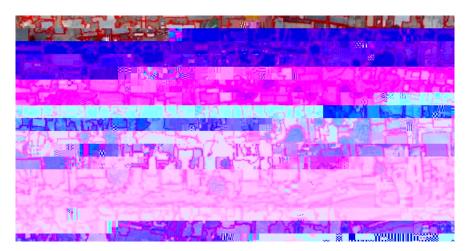
DAR ES SALAAM

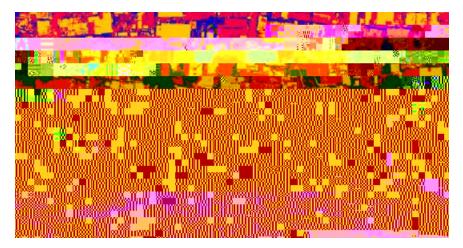
ZANZIBAR







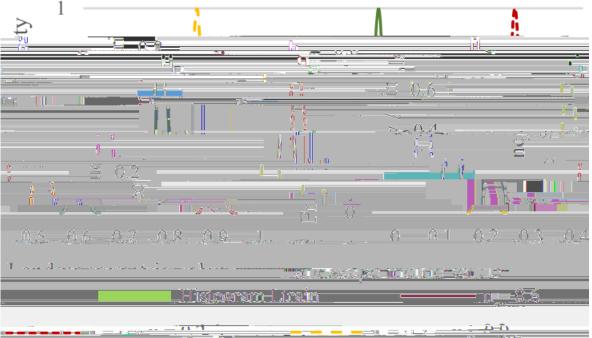






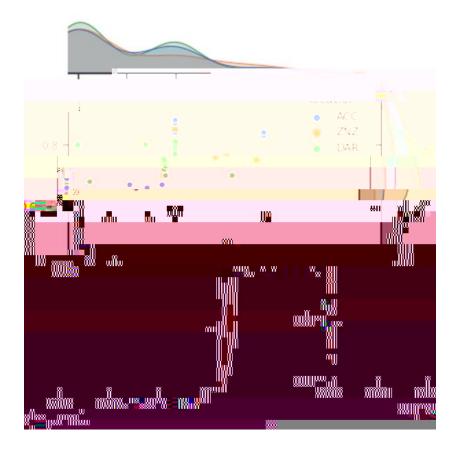
SIMILARITY SCORE

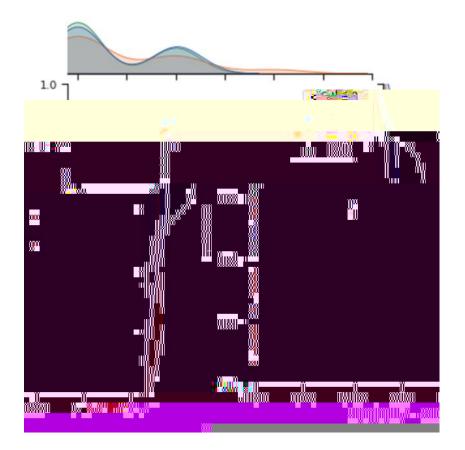














CONCLUSIONS

Biases challenge is identifying the sensitive attributes

methods to predict generalizability

