

Active Learning under Label Shift

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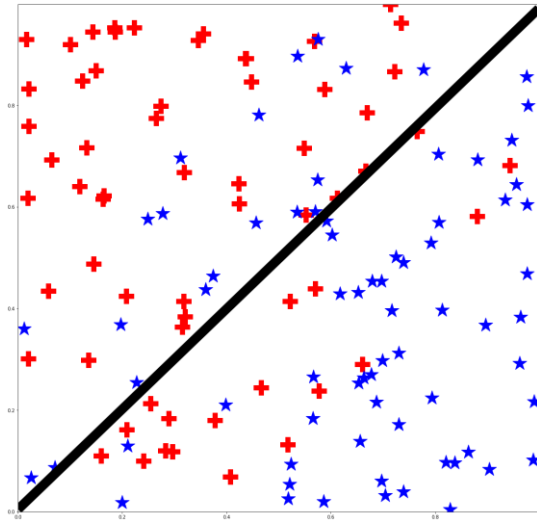
Caltech

The Label Shift Problem

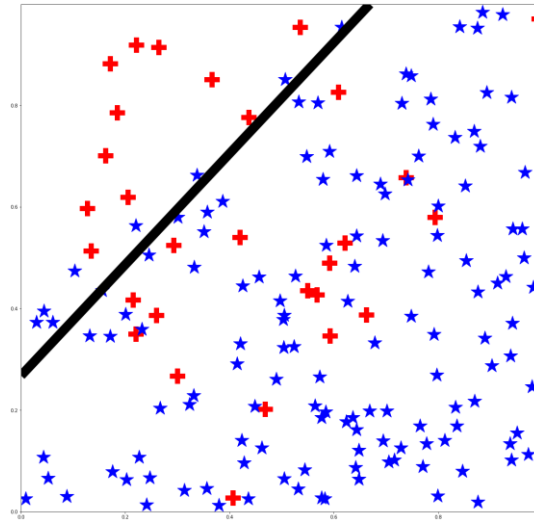
How to efficiently label despite distribution shift?

Label shift: certain classes are over/under-represented.

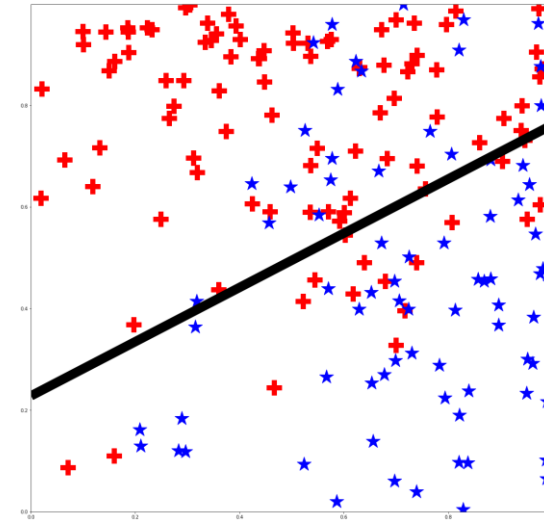
$$P(X | Y) = Q(X | Y) \text{ but } P(Y) \neq Q(Y)$$



Uniform Distribution



Label Shift



Covariate Shift




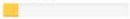








The Label Shift Problem

Label shift may arise from:

Societal bias

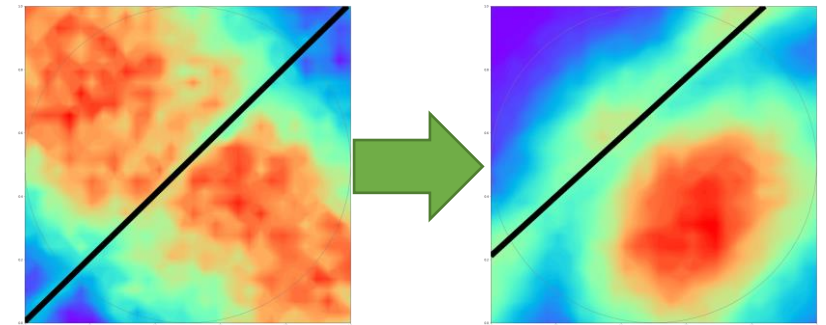
Class imbalance

Distorts uncertainty and disagreement measures.

Gender Classifier	Darker Subjects Accuracy	Lighter Subjects Accuracy	Error Rate Diff.
 Microsoft	87.1% 	99.3% 	12.2% 
 FACE++	83.5% 	95.3% 	11.8% 
 IBM	77.6% 	96.8% 	19.2% 



Source: Gender Shades (Buolamwini, Gebru)



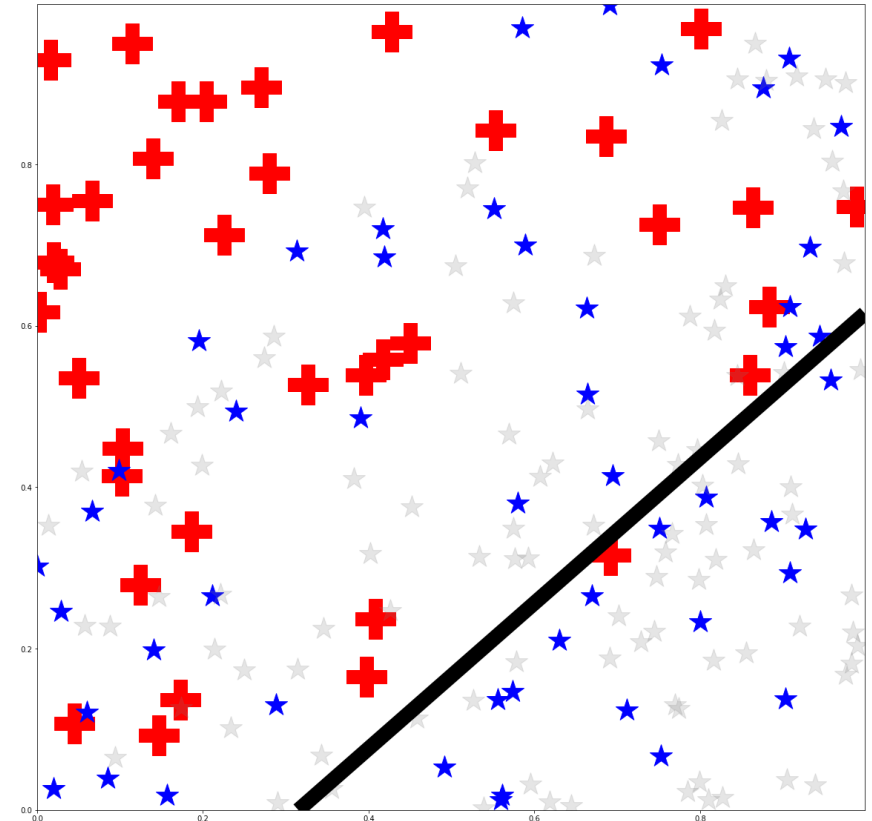
Change in sampling probability (DHM)
due to label shift.

Active Learning under Label Shift

Option 1: subsample by class

Option 2: importance weighting

ALLS: do a little of both



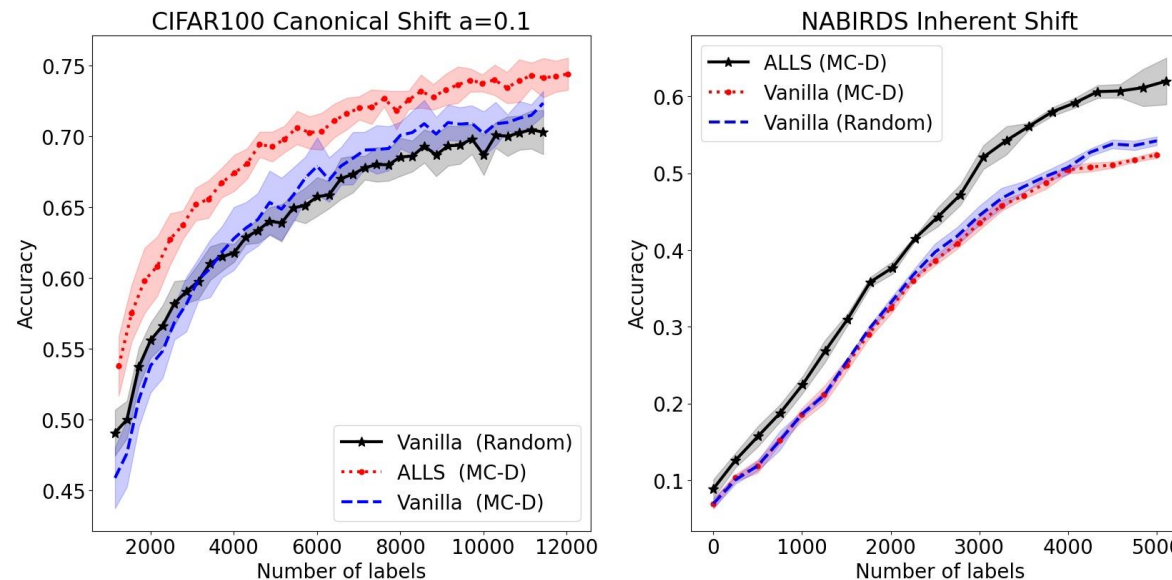
ALLS boosting red-class representation

Main Take-Aways

Preserves **consistency guarantee** with IW.

Reveals **bias-variance trade-off** between IW and subsampling.

Outperforms/matches with all explored settings/datasets.



...and **scales** to deep learning models and large output spaces