

# Context-Aware Drift Detection

Oliver Cobb and Arnaud Van Looveren

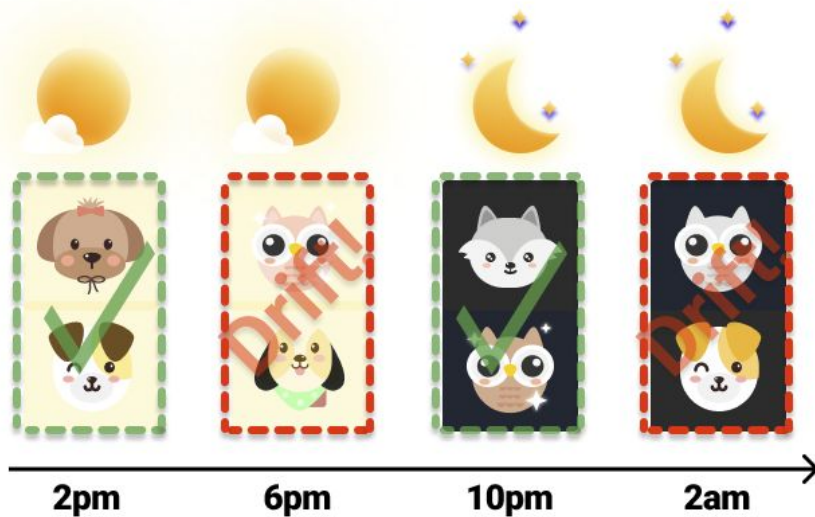


# Motivation

Training data

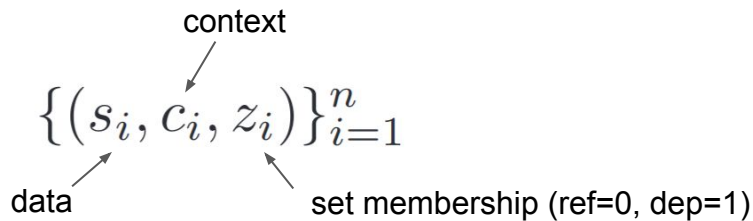


Deployment data



# Contextualised Samples

- Contextualised sample:



- Generative process:

$$\begin{aligned} Z &\sim P_Z, \\ C &\sim P_{C|Z}, \\ S^0 &\sim P_{S^0|C}, \quad S^1 \sim P_{S^1|C} \\ S &= S^0(1 - Z) + S^1 Z. \end{aligned}$$

# The Framework

- Null hypothesis:

$$h_0 : P_{S^0|C=c}(\cdot) = P_{S^1|C=c}(\cdot) \quad P_{C_1}\text{-a.e.}$$

- Test statistic:

$$E[U_D(C)|Z = 1] \quad \text{where} \quad U_D(c) = D(P_{S^0|C=c}, P_{S^1|C=c})$$

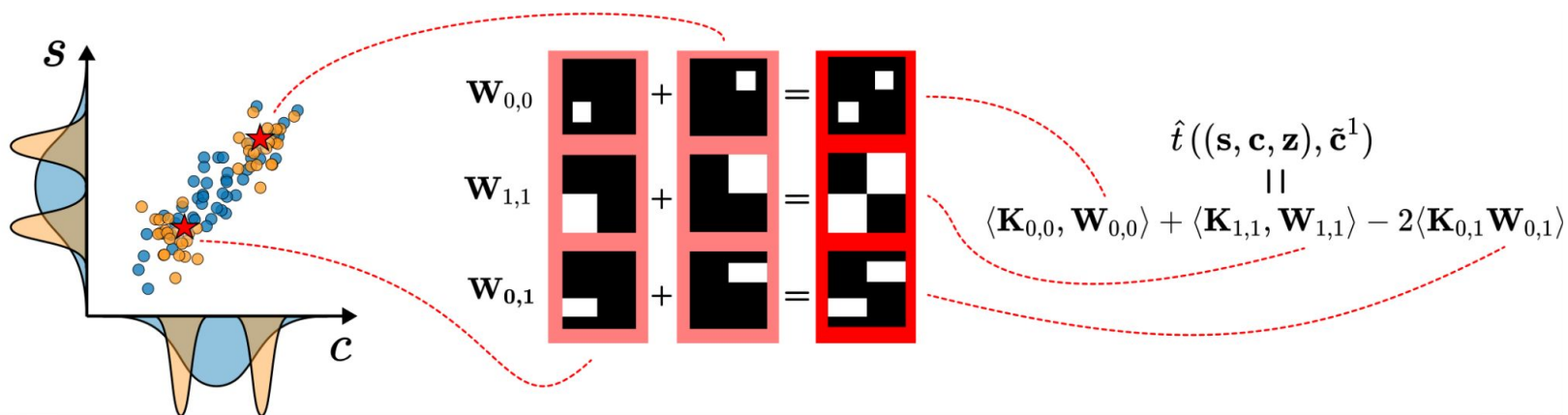
ADiTT

CoDiTE

- p-value: conditional permutation test

# MMD-ADiTT

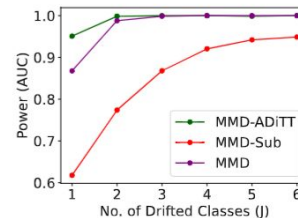
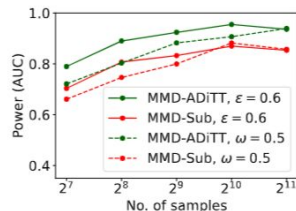
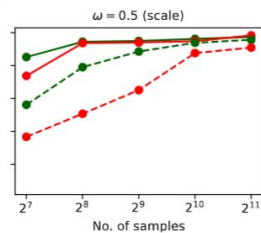
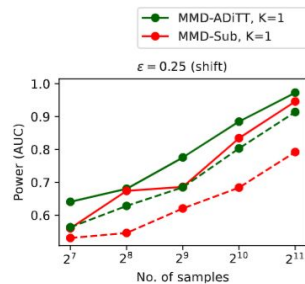
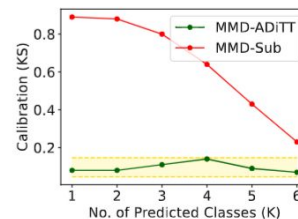
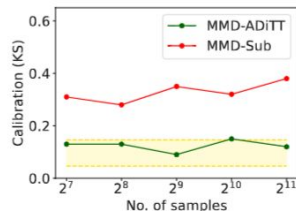
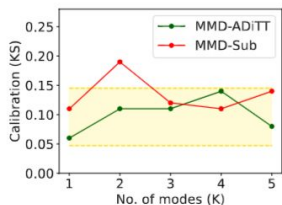
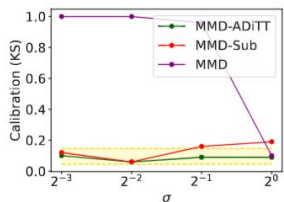
- MMD-based CoDiTE developed by Park et al. (ICML 2021)<sup>1</sup>.
- We introduce a corresponding ADiTT estimator



<sup>1</sup> Conditional distributional treatment effect with kernel conditional mean embeddings and u-statistic regression

# Further Applications and Results

- May alternatively wish to allow variation in:
  - Subpopulation prevalences
  - Model predictions
  - Model uncertainty



Thanks for watching!



[github.com/SeldonIO/alibi-detect](https://github.com/SeldonIO/alibi-detect)