

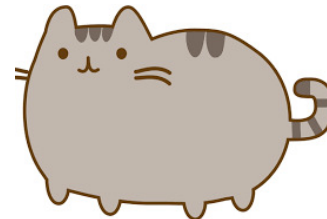


Covariate shift

Training set



At test time





Why would anyone be so stupid?

Covariate Shift



- **Web search**
 - Training - page relevance data for the US market
 - Testing - recommend pages for Canada (UK, Australia)
- **Speech recognition**
 - Training - West coast accent
 - Testing - Southern drawl, Texan, non-native speaker
- **Language**
 - Training - 'James, bring me a **soda**'
 - Testing - 'John, bring me a '**pop**' (or coke, etc.)

Covariate Shift



- **Medical**

- Training - University students + old men with prostate cancer
- Testing - Potentially sick old men

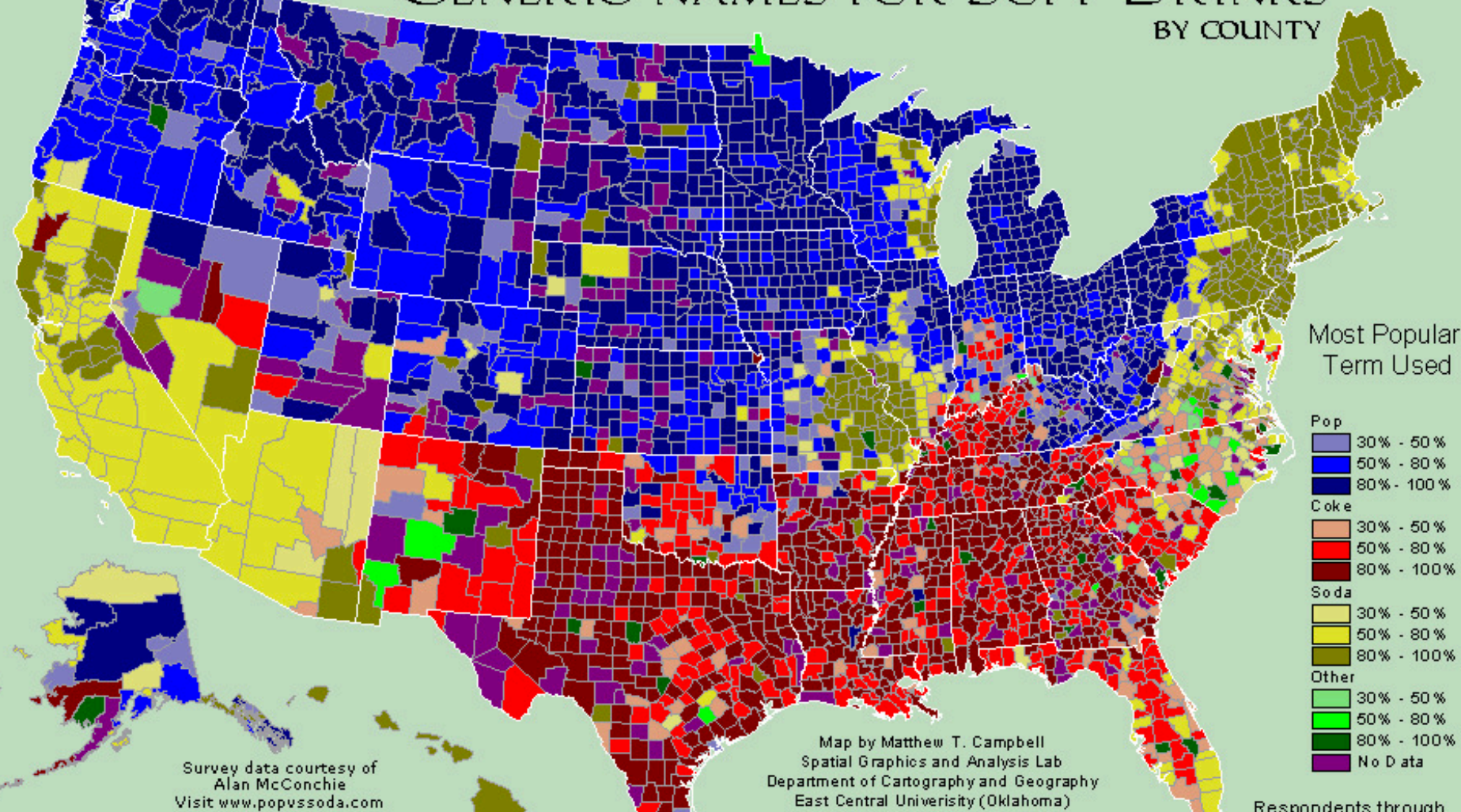
- **Reinforcement Learning**

- Training - Data gathered with current policy
- Testing - Environment reacting to updated policy

- **Databases**

- Training - DB tuned to 2017 usage pattern
- Testing - DB deployed on AWS in 2018

GENERIC NAMES FOR SOFT DRINKS BY COUNTY



Real Story from CES'19

- Startup with smart vending machine demo that identifies purchases via a camera
- Demo at CES failed
 - Different light temperature
 - Light reflection from table
- The fix
 - Collect new data
 - Buy tablecloth
 - Retrain all night



What is happening?

$$q(x, y) = q(x)p(y|x)$$



- Training Risk

Training data

$$\underset{w}{\text{minimize}} \int dx p(x) \int dy p(y|x) l(f(x, w), y)$$

or rather $\underset{w}{\text{minimize}} \frac{1}{m} \sum_{I=1}^m l(f(x_i, w), y_i)$

- Test Risk is different

Test data


$$\int dx q(x) \int dy p(y|x) l(f(x, w), y)$$



Fixing it (covariate shift correction)



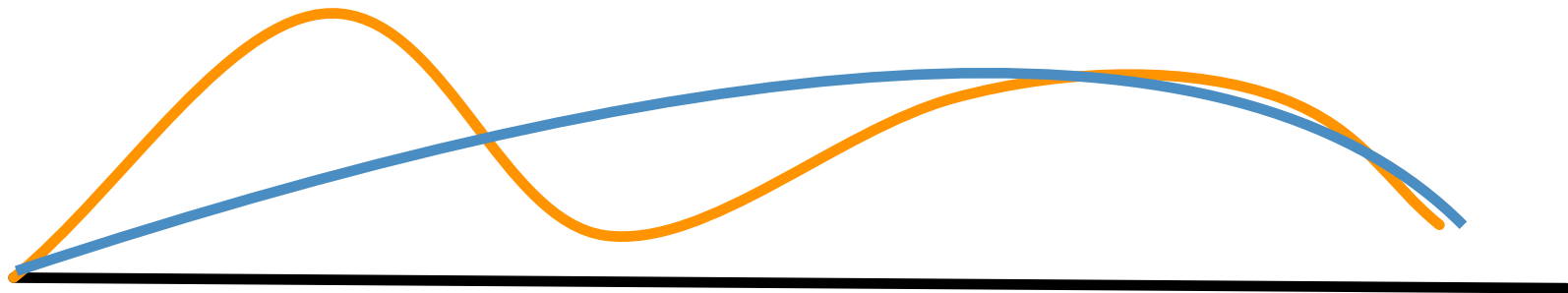
- Density ratio

$$\int dx q(x) f(x) = \int dx p(x) \underbrace{\frac{q(x)}{p(x)}}_{\alpha(x)} f(x) = \int dx p(x) \alpha(x) f(x)$$


- Need to find density ratio, but we don't have p or q .
- Estimating densities is hard.
E.g. what to do if density is very high (or very low)?

$$r(x, y) = \frac{1}{2} [p(x) \cdot \delta(y, 1) + q(x) \cdot \delta(y, -1)]$$

Density Ratio Estimation



- If $p(x)$ and $q(x)$ differ, we can train a classifier to distinguish between them.

- Classification

$$r(x, y) = \frac{1}{2} [p(x) \cdot \delta(y, 1) + q(x) \cdot \delta(y, -1)]$$

Fixing it (covariate shift correction)



$$r(x, y) = \frac{1}{2} [p(x) \cdot \delta(y, 1) + q(x) \cdot \delta(y, -1)]$$

- Conditional class probability

$$r(y = 1|x) = \frac{p(x)}{p(x) + q(x)} \text{ and hence } \alpha = \frac{q(x)}{p(x)} = \frac{r(y = -1|x)}{r(y = 1|x)}$$

- Simple algorithm

- Train classifier between training and test set

- Reweight training data with $\frac{r(y = -1 | x)}{r(y = 1 | x)}$

- Can use Generalization Performance Estimate to decide whether we even have Covariate Shift.

Corrected Training Problem



- Original Problem

$$\underset{w}{\text{minimize}} \sum_{i=1}^m \frac{1}{m} l(y_i, f(x_i, w))$$

- Reweighted Problem

$$\underset{w}{\text{minimize}} \sum_{i=1}^m \alpha_i l(y_i, f(x_i, w)) \text{ where } \alpha_i = \frac{r(y = -1 | x_i)}{r(y = 1 | x_i)}$$

- Problems

- Coefficients α_i are estimates. Adds variance & bias.
- What if p and q are very different?

Problems with p and q



$$\sum_{i=1}^m \alpha_i l(y_i, f(x_i, w))$$

Problems with p and q



$$\sum_{i=1}^m \alpha_i l(y_i, f(x_i, w))$$

- Effective sample size

$$m^* = \frac{\|\alpha\|_1^2}{\|\alpha\|_2^2}$$

- Unmodified weights
 - Less bias
 - High variance for small m^*
- Clipped weights $\bar{\alpha}_i = \min(\alpha_i, C)$
 - Potentially some bias
 - Smaller variance
(larger effective sample size)

Key Takeaways



Covariate Shift assumes

$$p(x, y) = p(x)p(y | x) \longrightarrow q(x, y) = q(x)p(y | x)$$

- Label dependency for training and test is the same
- Covariate distribution is different

Detection & Fix

- Train binary classifier between training and test set
- If nontrivial accuracy, use it to reweight training data via
$$\alpha(x) = r(y = 1 | x) / r(y = -1 | x)$$
- Fix effective sample size by clipping $\bar{\alpha}(x) = \min(\alpha(x), C)$