Machine Learning with Opponents

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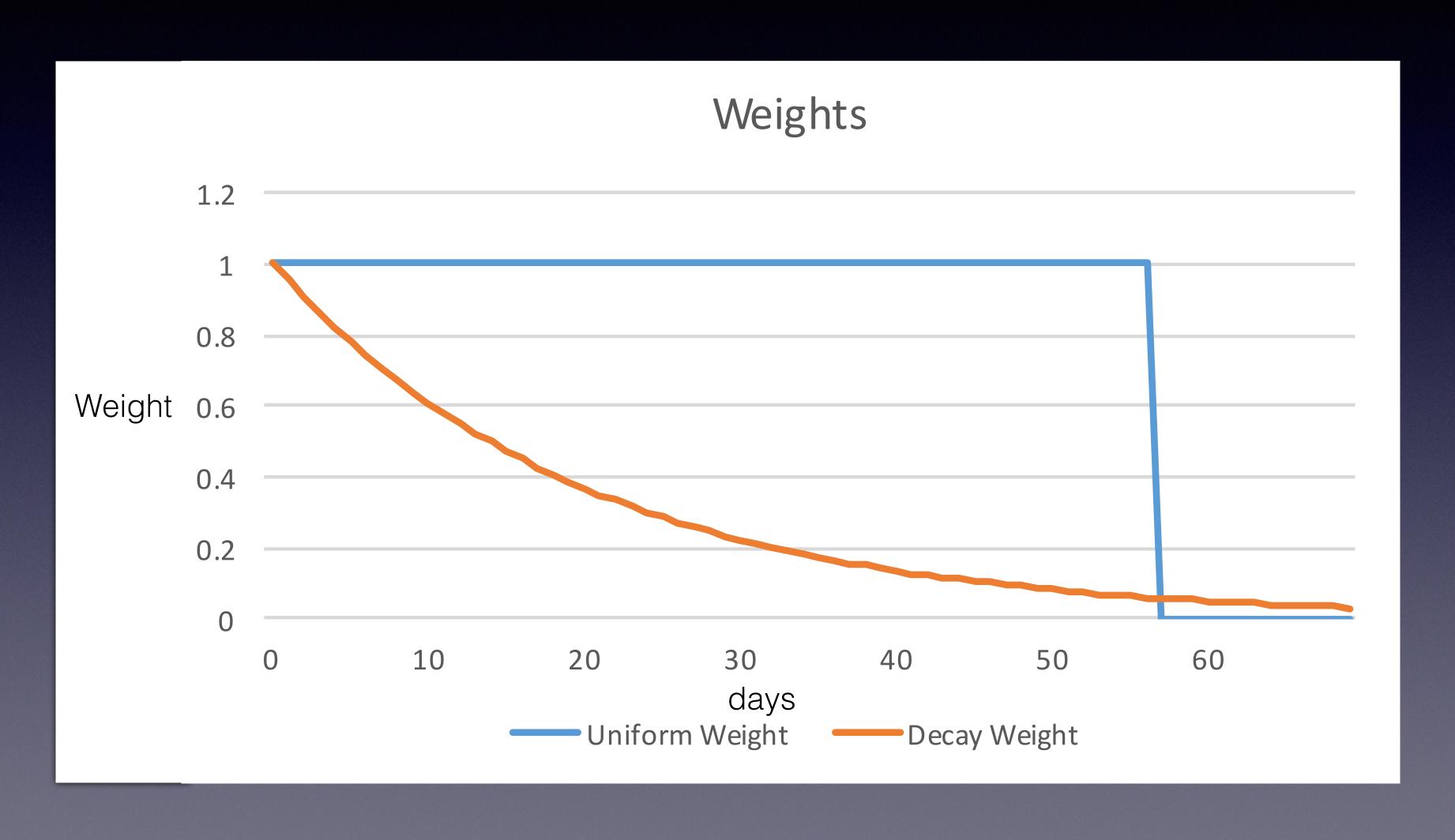
Observations Feature engineering Models

Observations

Observation Weighting

- Re-weight observations, based on:
 - Uniform weight
 - Observation age (staleness)
 - Random down-sampling (randomly set weight to 0)
 - Up-sampling known opponent attacks (raise weight for known attacks)

Observation Weighting



(Synthetic Minority Over-sampling Technique)

- Create synthetic observations, similar to actual attacks
- Goal: Better model rare events (opponent attacks)
- · Majority class: Down sample, with some probability
- Minority class: Create 'synthetic' observations

- 1. Select minority point
- 2. Select neighbor
- 3. Create new point

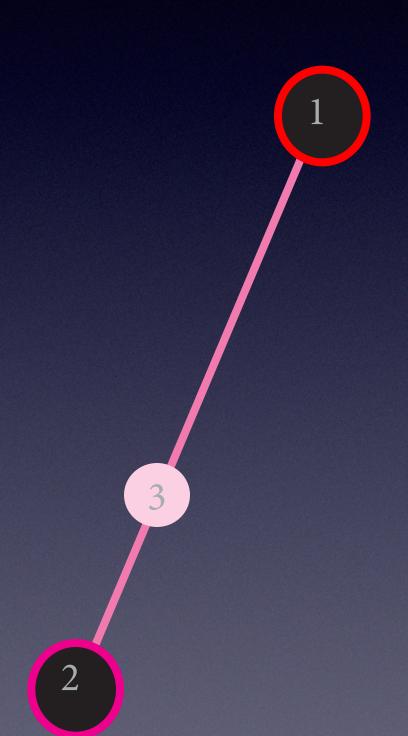
- 1. Select minority point
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Observation Weighting SMOTE Sampling

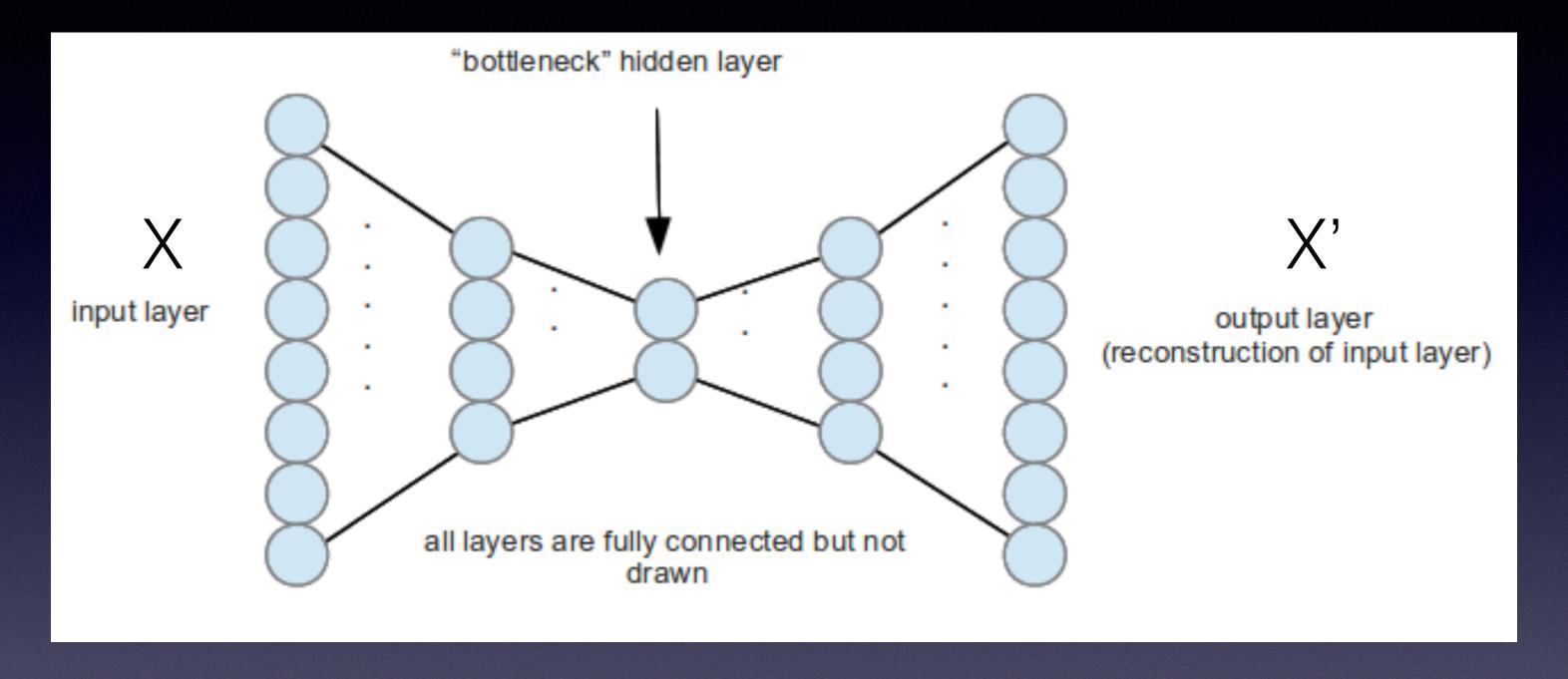


Feature engineering

Outlier Detection

- Goal: Create outlier score
- Method: Train learner to re-create input vector
 - PCA: Reduce dimensionality, increase dimensionality
 - Neural Network: Train auto-encoder, with bottleneck
- Score: Measure distance from output vector to input vector

Outlier Detection



Outlier score: |X'-X|

Microclustering

- Goal: Capture what is 'normal' for this user
- Method:
 - Cluster users into small cohorts of similar users
 - Train a model for each cohort
- Score: Fraud likelihood score for user's cohort

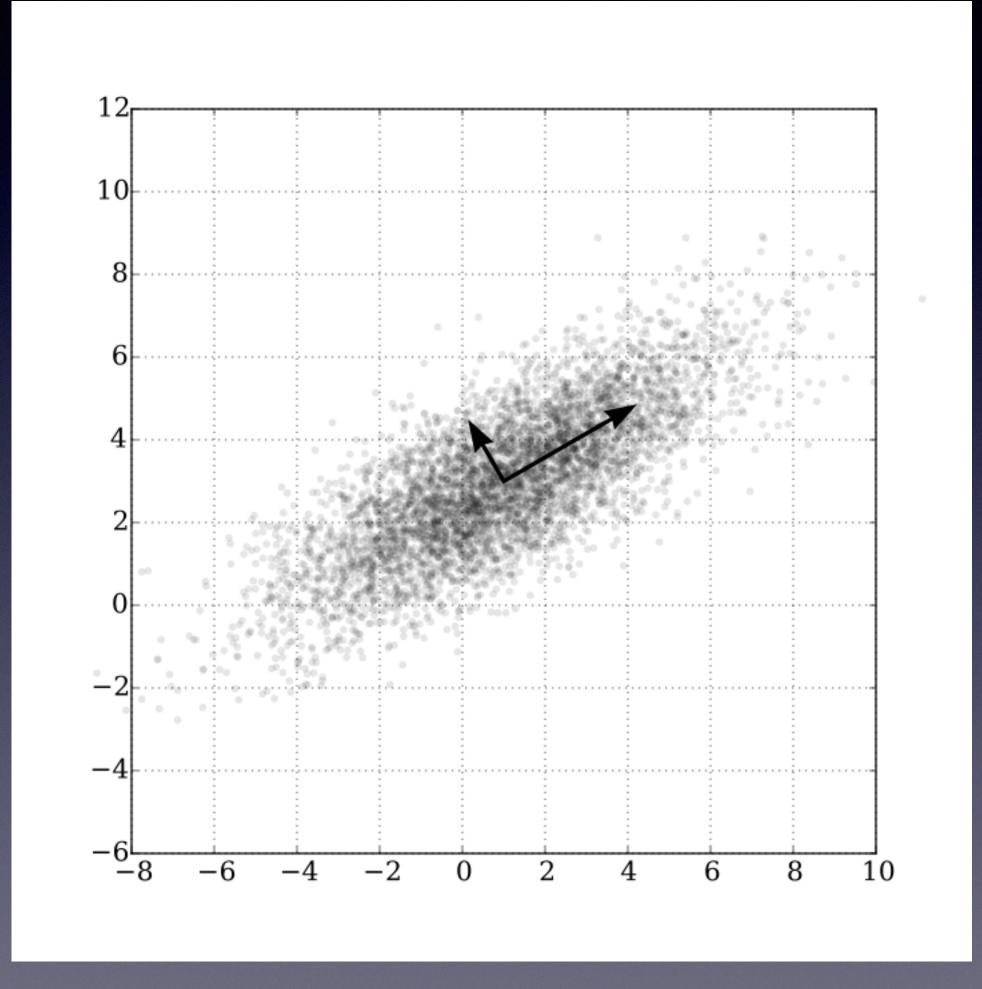
Historical aggregates

- Goal: Capture what is 'normal' for this user
- Method: Perform aggregates for various time windows
 - E.g. Average spend for past 30, 60, 90 days
 - E.g. Maximum spend for past 30, 60, 90 days

Low Rank Models

- Goal: Reduce dimensionality, to include sparse or previously unused variables
- Method: Reduce dimensionality with generalized PCA
 - Model directly on components (latent factors)

Low Rank Models



Outlier Detection
Microclustering
Historical aggregates
GLRM



Models

Grid search

- Goal: Find optimal hyper-parameters for given class of models
- Method: Create every possible permutation of hyper-parameters, and compute models until heat death of universe

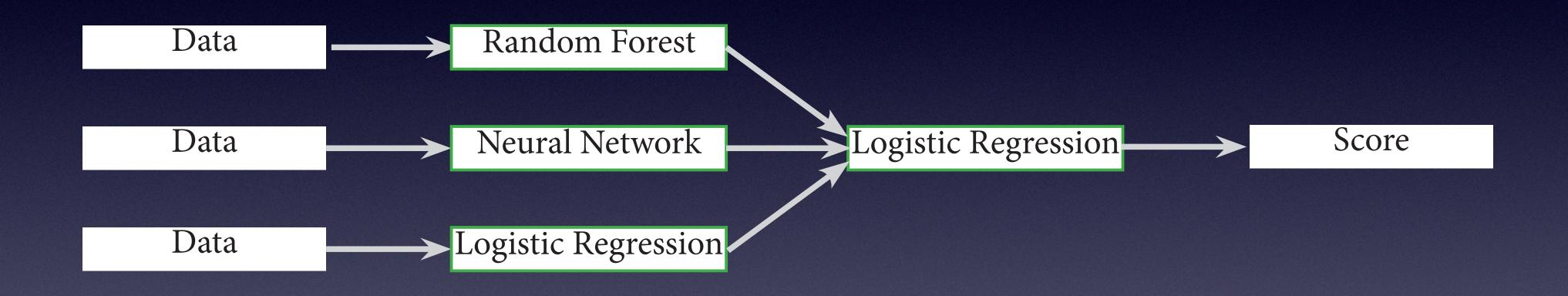
Neural Networks

- Goal: Deep, non-linear models perform well with rare cases
- Method: Try many different architectures, with many different hyper parameters
 - Feed previous user transactions to Bidirectional LSTM
 - Auto-encoder

Ensemble Modeling

- Goal: Leverage a diverse set of algorithms
- Method: Train multiple classes of algorithms (tree based, linear, neural network), possibly with multiple hyper-parameters. Combine scores with meta model (such as Logistic Regression w/ nonnegative betas)

Ensemble Modeling



Grid Search Neural Networks Ensemble models



Thanks!

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