

# Machine Learning with Opponents

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Slides: <https://goo.gl/wY8iFY>

MAN, I SUCK AT THIS GAME.  
CAN YOU GIVE ME  
A FEW POINTERS?

0x3A28213A  
0x6339392C,  
0x7363682E.

I HATE YOU.





# Overview

Sampling  
Feature Engineering  
Modeling

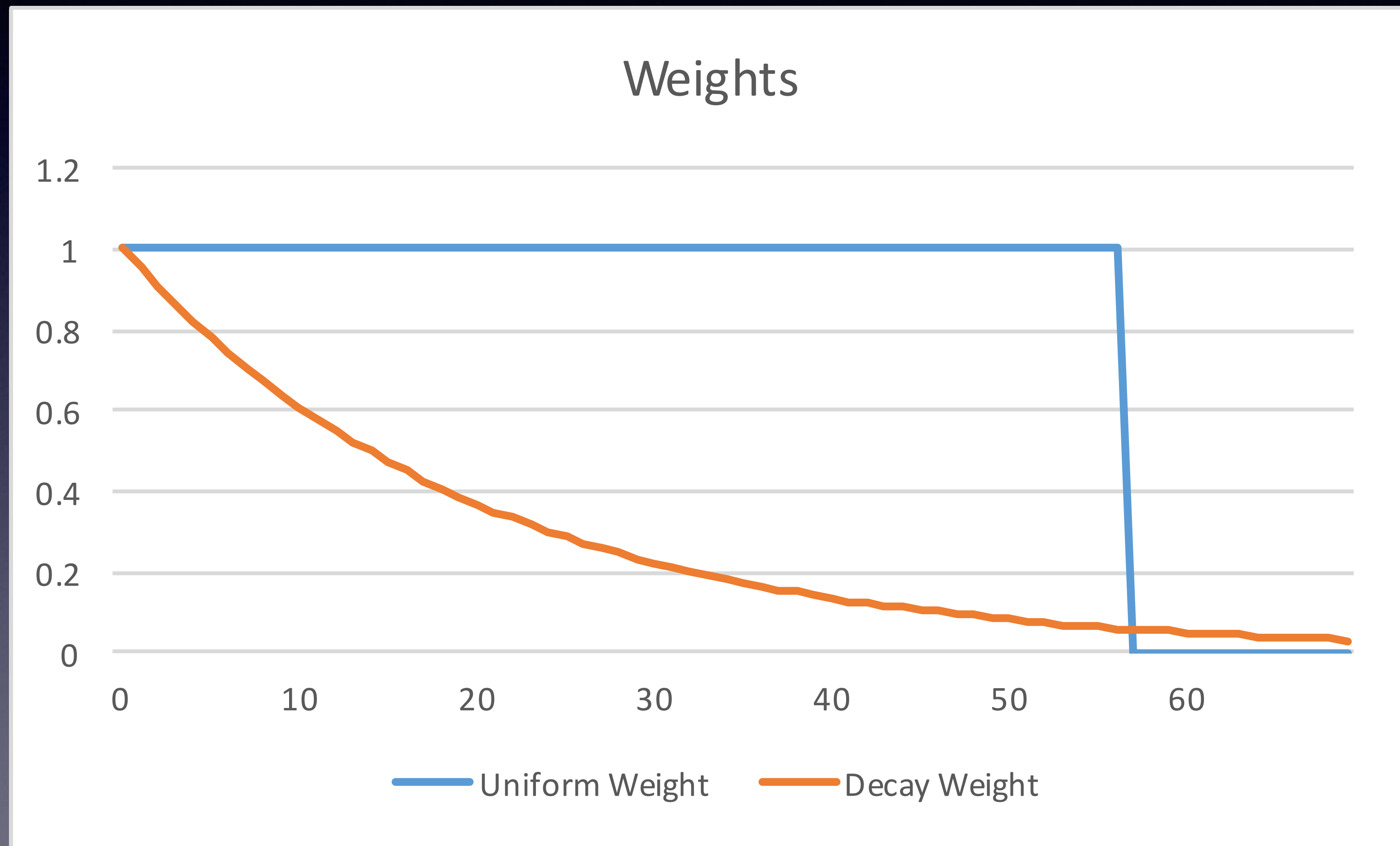


# Sampling

# Observation Weighting

- Effect cost function by weighting every row at train time
- Weights include
  - Uniform weight
  - Observation age (staleness)
  - Random down-sampling
  - Up-sampling known opponent attacks

# Observation Weighting



# SMOTE

(Synthetic Minority Over-sampling Technique)

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



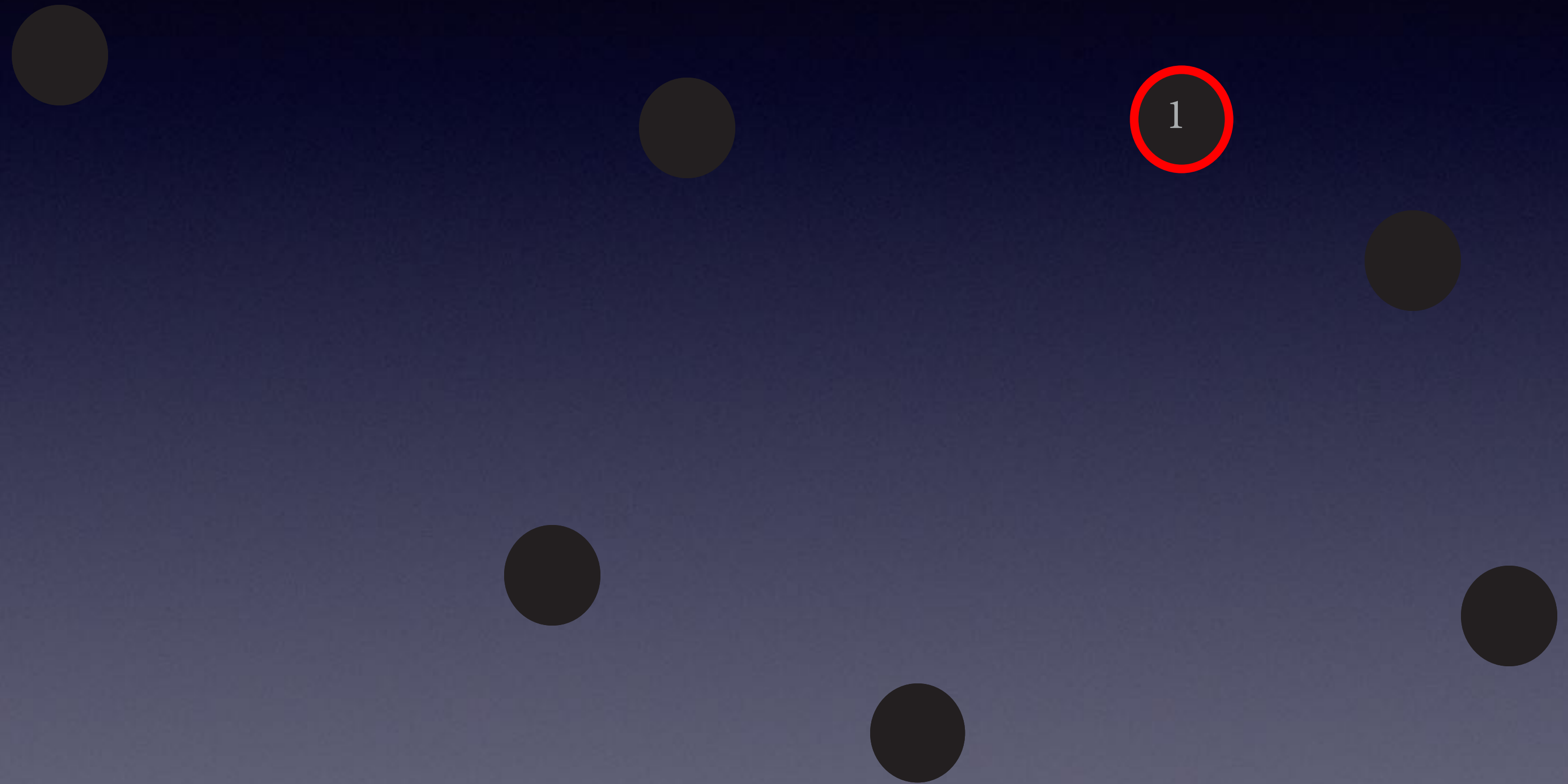
# SMOTE

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



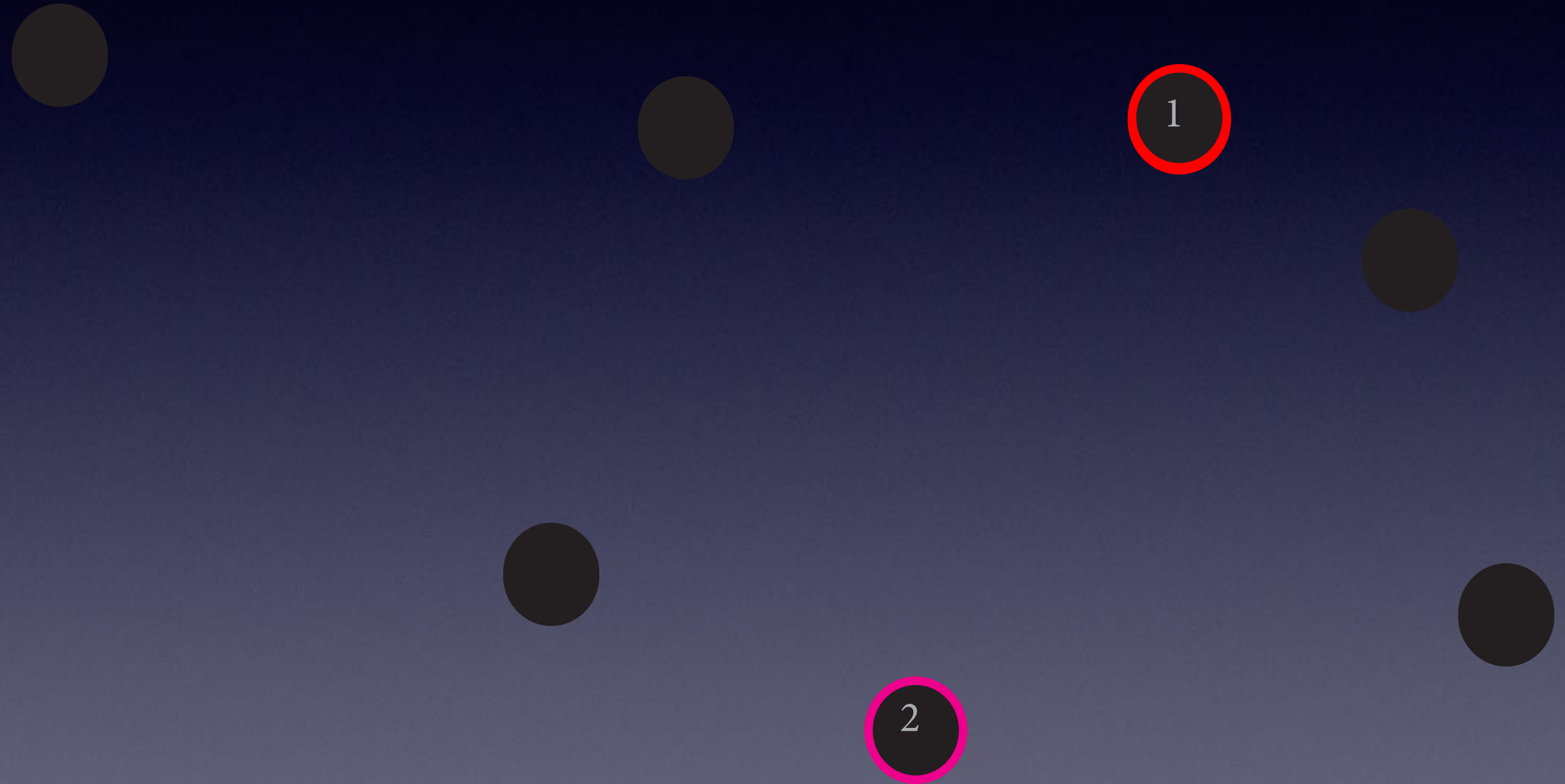
# SMOTE

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



# SMOTE

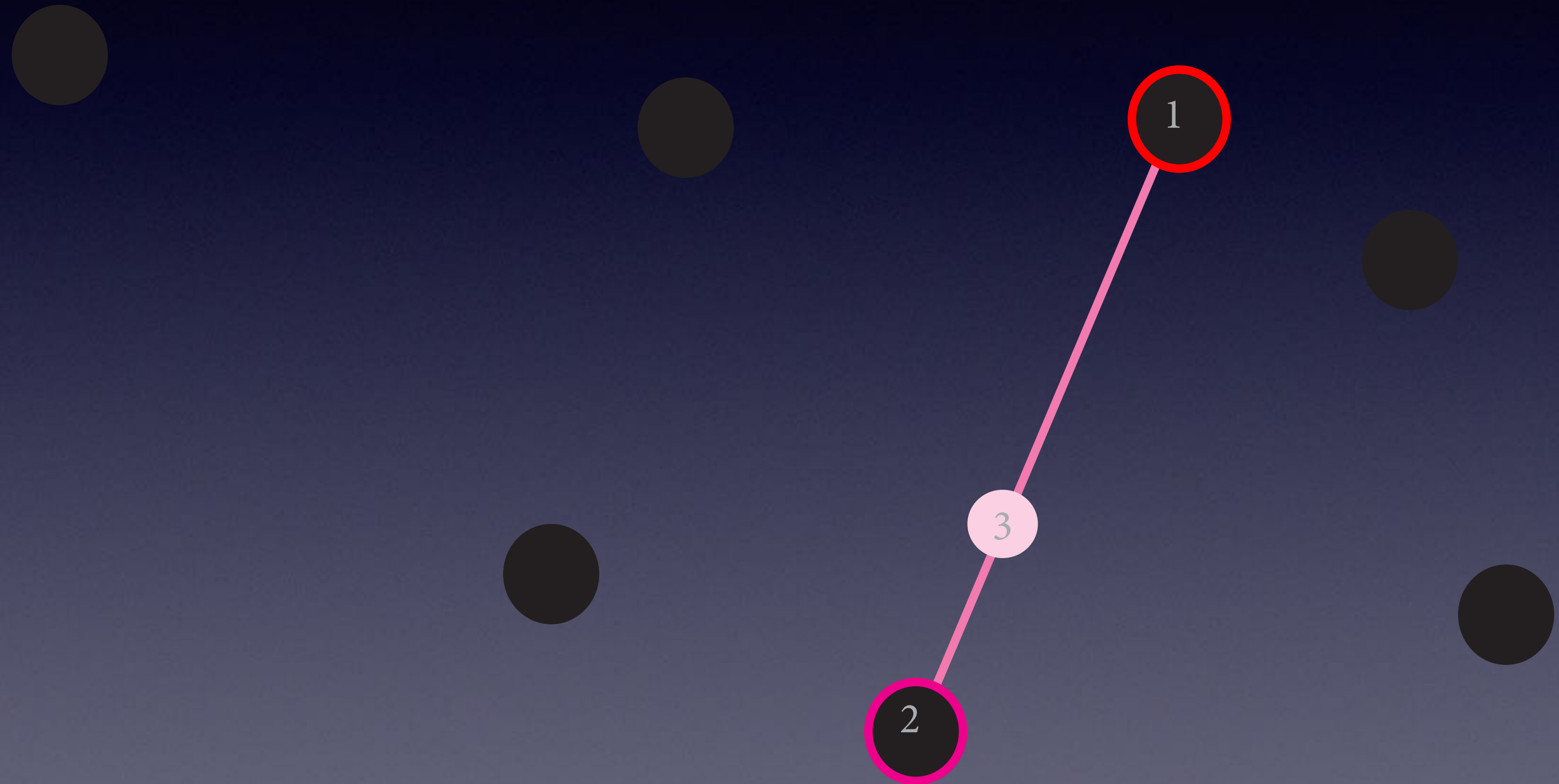
1. Select minority point
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# SMOTE

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# Observation Weighting SMOTE Sampling



# Features

# Outlier Detection

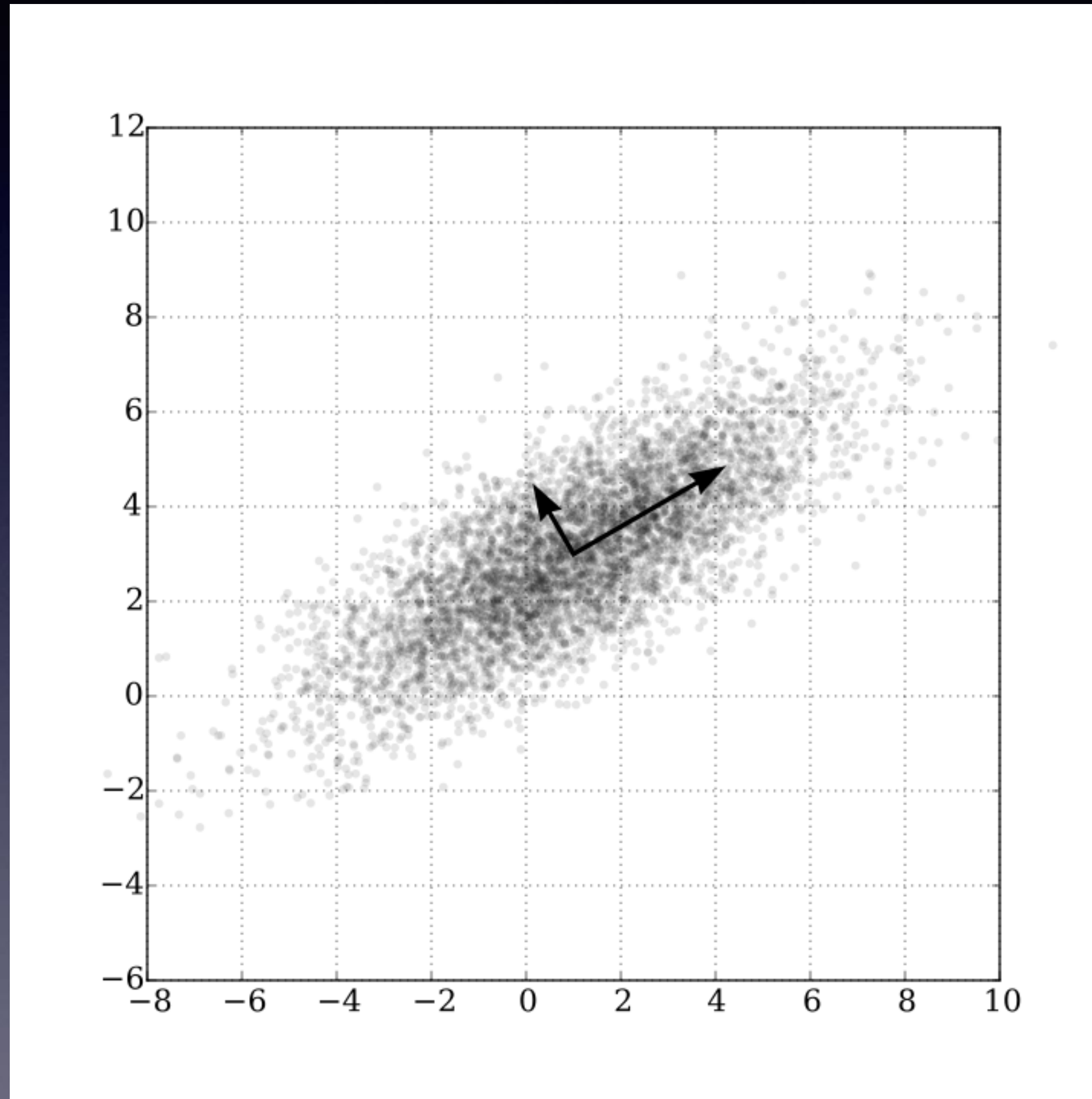
- Goal: Create outlier score
- Train learner to re-create input vector
  - PCA: Reduce dimensionality, increase dimensionality
  - Neural Network: Train auto-encoder, with bottleneck
- Measure distance from output vector to input vector



# Low Rank Models

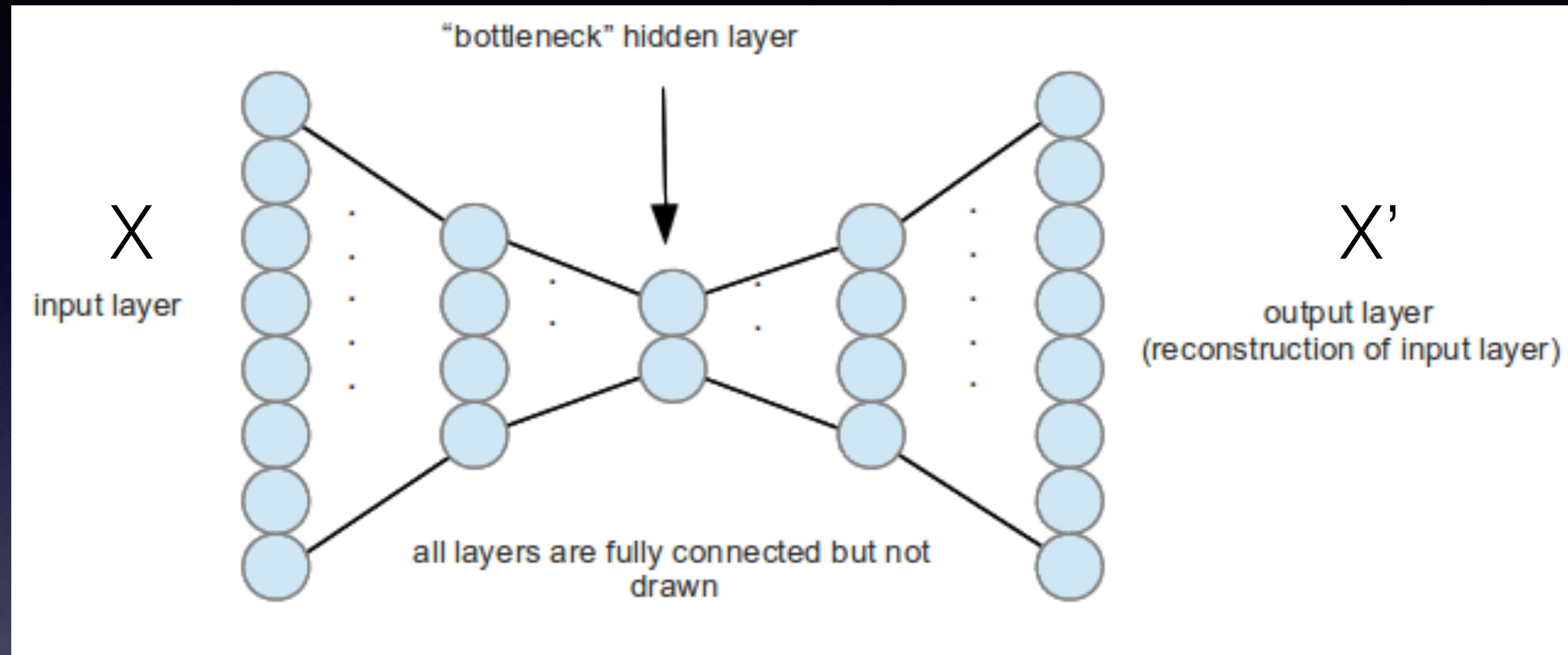
- Goal: Reduce dimensionality for dataset with many variables
- Reduce dimensionality with generalized PCA
- Model directly on components (latent factors)

# Low Rank Models



<https://web.stanford.edu/~boyd/papers/pdf/qlrm.pdf>  
<https://github.com/h2oai/h2o-tutorials/blob/master/tutorials/qlrm/qlrm-tutorial.md>

# Outlier Detection



Outlier score:  $|X' - X|$



# Outlier Detection

## GLRM





# Modeling

# Grid search

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

# Neural Networks

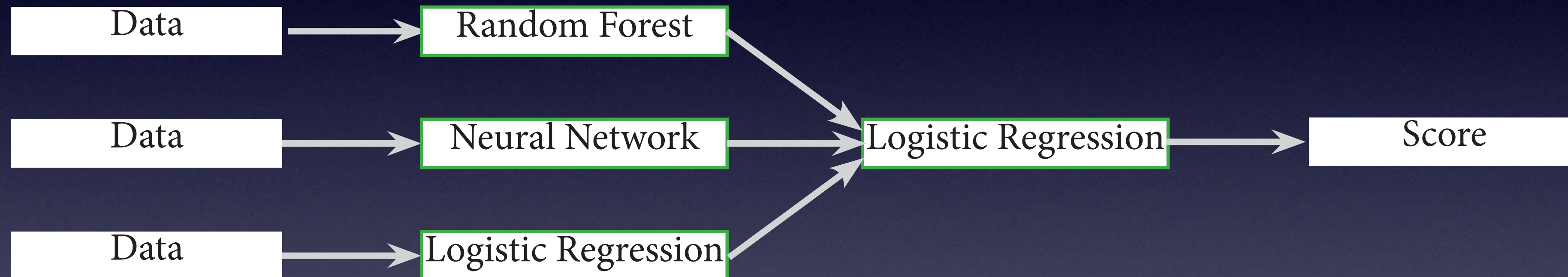
- Too complicated to cover here 😞
- Strengths: Able to capture complicated, non-linear relationships. Deals well with class imbalance, near-arbitrary data types
- Weaknesses: Difficult to train, many free parameters



# Ensemble Modeling

- Goal: Leverage a diverse set of algorithms
- Train multiple classes of algorithms (tree based, linear, neural network), possibly with multiple hyper-parameters, combine scores with meta model

# Ensemble Modeling



Grid Search  
Neural Networks  
Ensemble models





Thanks!

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