# Implementing Bin Counting and Feature Hashing



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#### Overview

Converting continuous data into categorical data

Bucketing continuous data into bins

Bucketing data using Pandas and the KBinsDiscretizer

Hash nominal features to numeric features

## Types of Data

Categorical

Male/Female, Month of year

Numeric (Continuous)

Weight in lbs, Temperature in F

All other forms of data, such as text and image data, must be converted to one of these forms

# Bucketing

Categorical

Male/Female, Month of year

Numeric (Continuous)

Weight in lbs, Temperature in F

Bucketing techniques to convert continuous data to discrete categories

Categorical

Male/Female, Month of year

Numeric (Continuous)

Weight in lbs, Temperature in F

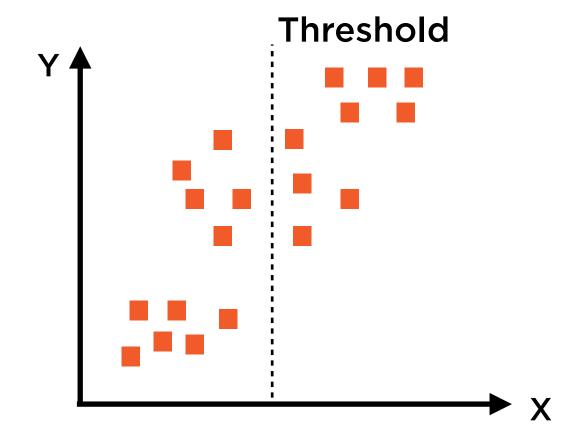
Converting data to numeric representations of lower dimensionality

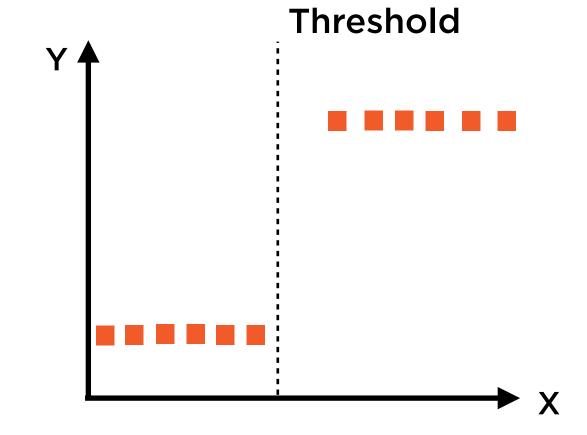
# Bucketing Continuous Data

# Binarizer

Converts continuous variable into a binary categorical variable based on a threshold specified by user.

## Binarizer





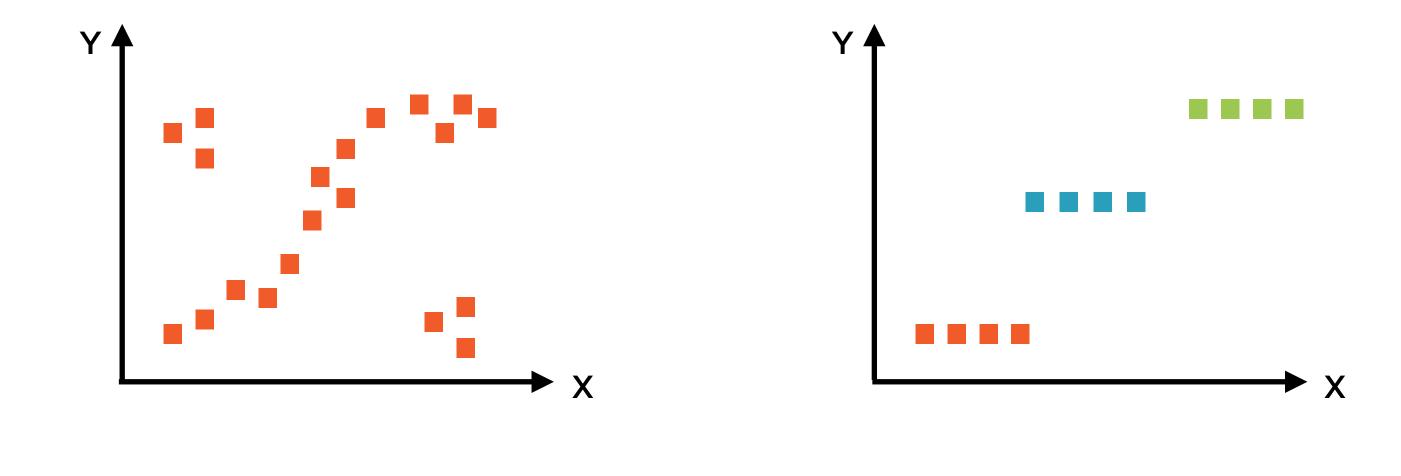
Continuous Input

Binary Categorical Output

# KBinsDiscretizer

Generalizes idea of binarizer; converts continuous data into categorical data arranged into a specified number of bins.

## KBinsDiscretizer



After (3 Bins)

**Before** 

## KBinsDiscretizer Strategies

#### Uniform

Bin widths are constant in each feature

#### Quantile

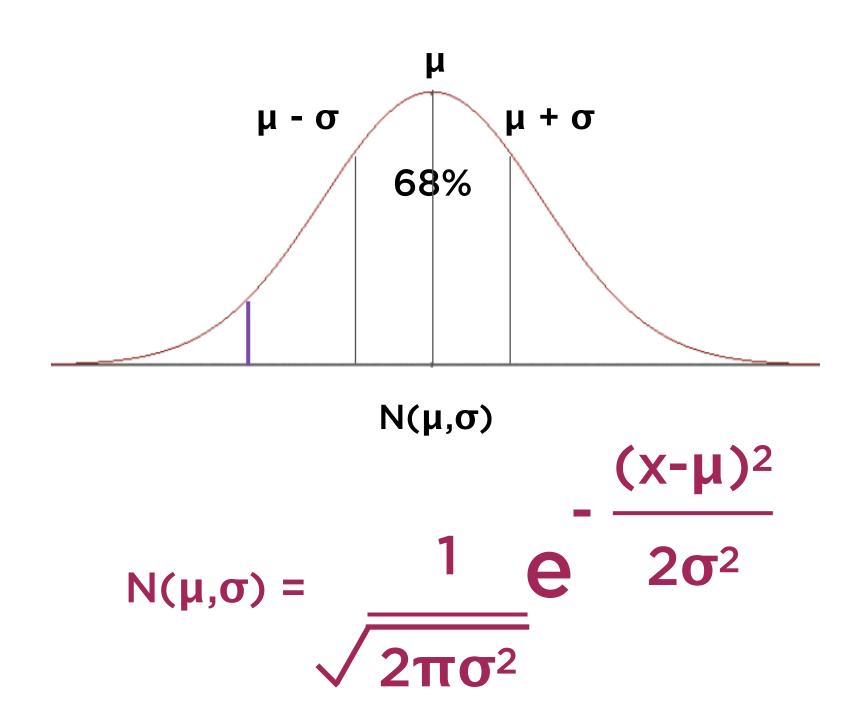
All bins in each feature have approximately the same number of samples

#### K-means

Bins based on the centroids of a K-means clustering procedure

# A graph showing the count of values in each bin is called a Histogram

### Continuous Distribution

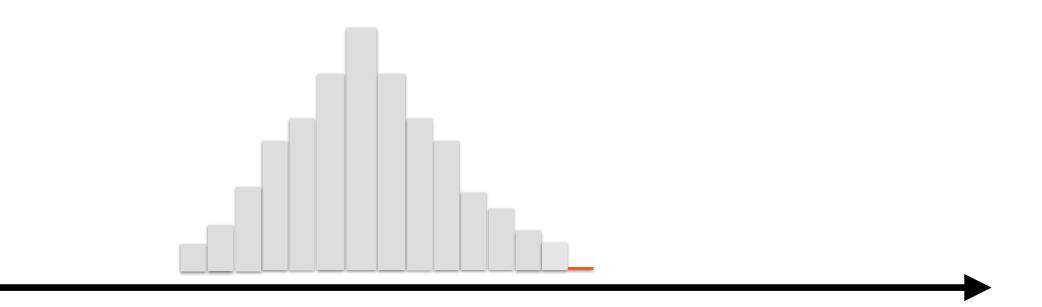


#### Data Drawn from Distribution

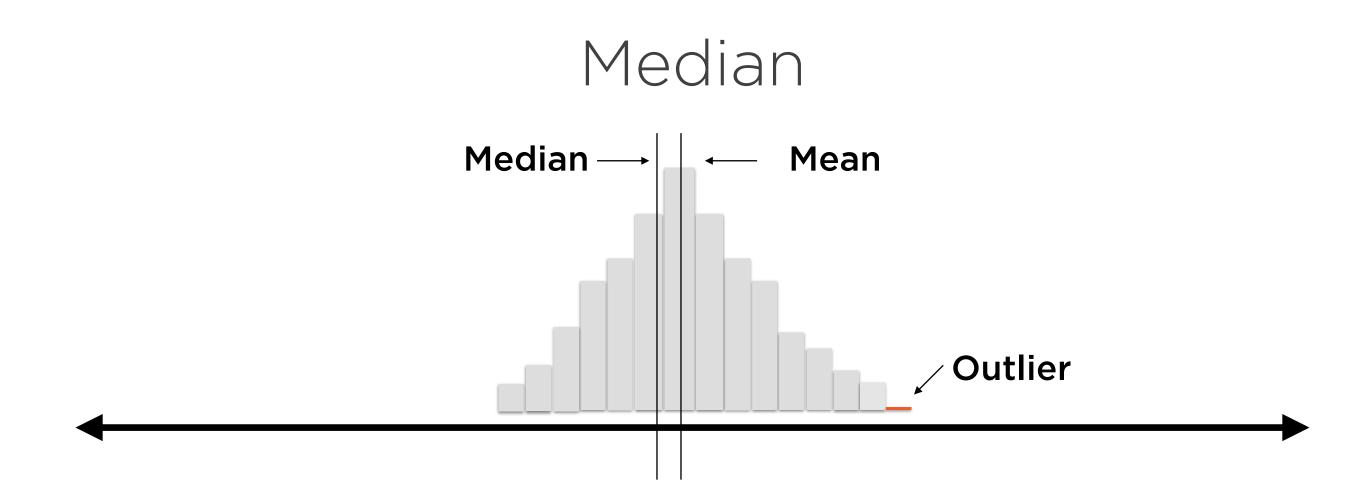


Outliers might represent data errors, or genuinely rare points legitimately in dataset

# Histogram of Bin Counts

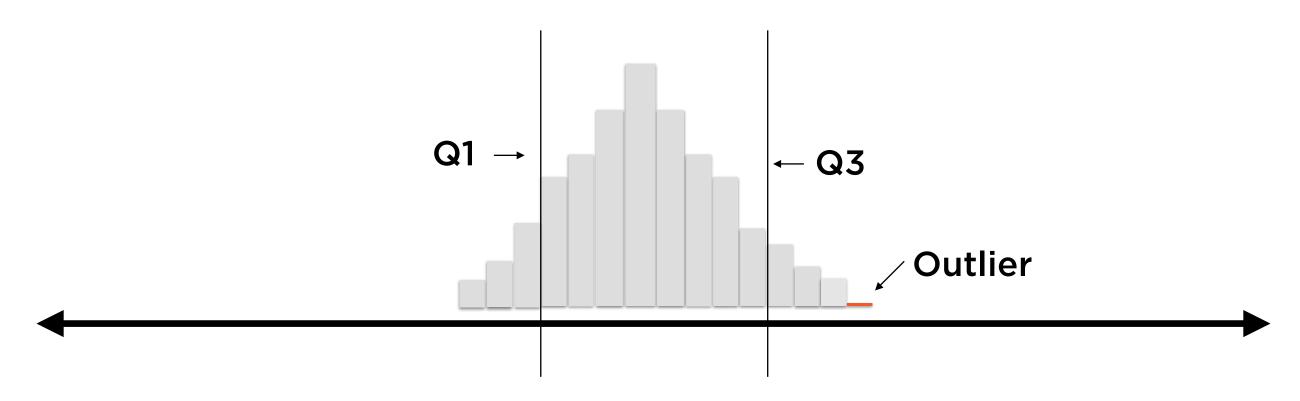


Bucketize data and count how many data points fall within each bucket



Median = 50th percentile: 50% of points on either side

# Histogram of Bin Counts



Q3 = 75th percentile: 75% of points smaller than this

Q1 = 25th percentile: 25% of points smaller than this

Inter-quartile Range (IQR) = 75th percentile - 25th percentile

## Demo

Bucketing continuous data using Pandas

## Demo

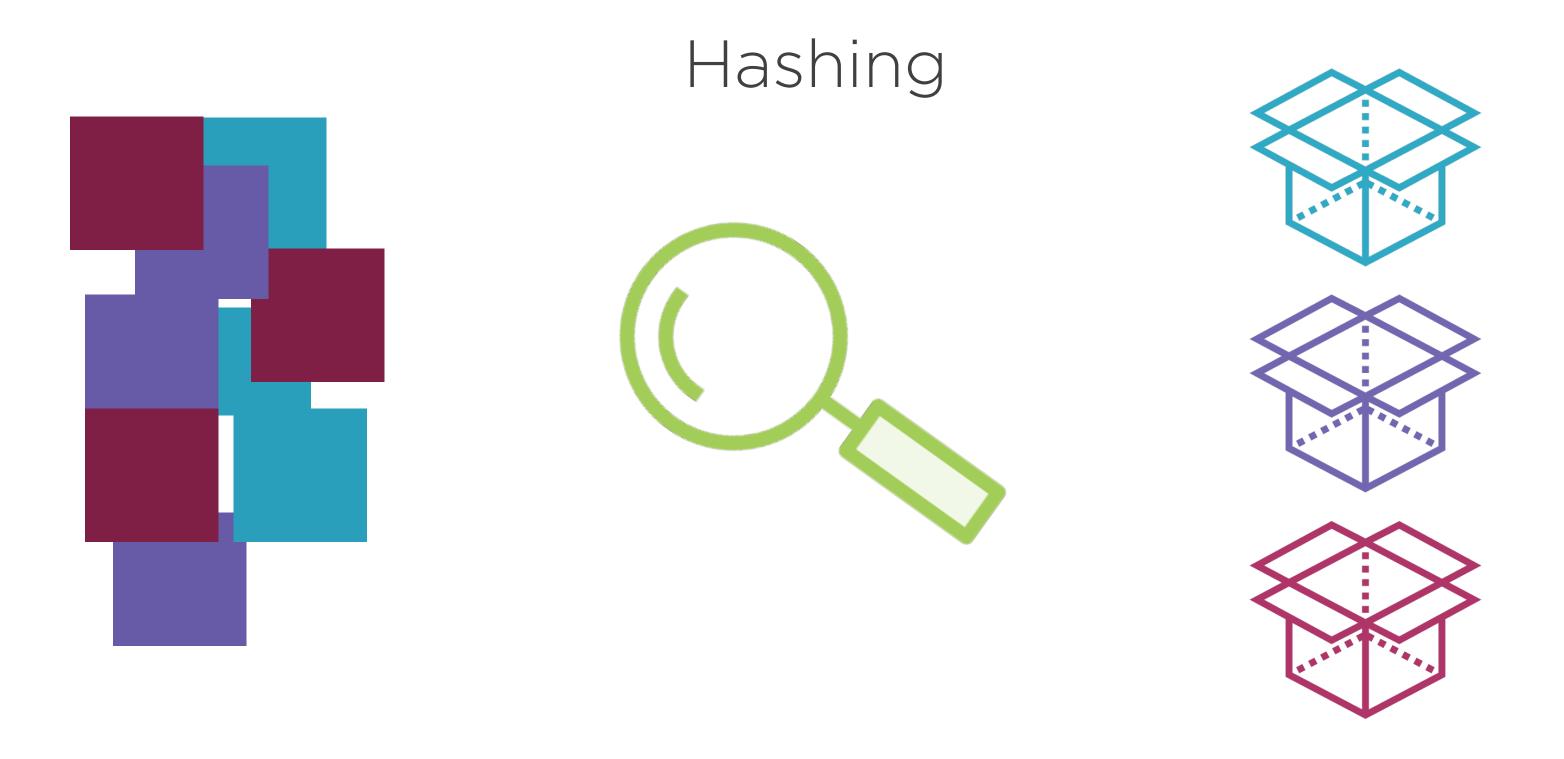
Discretizing continuous data using the KBinsDiscretizer



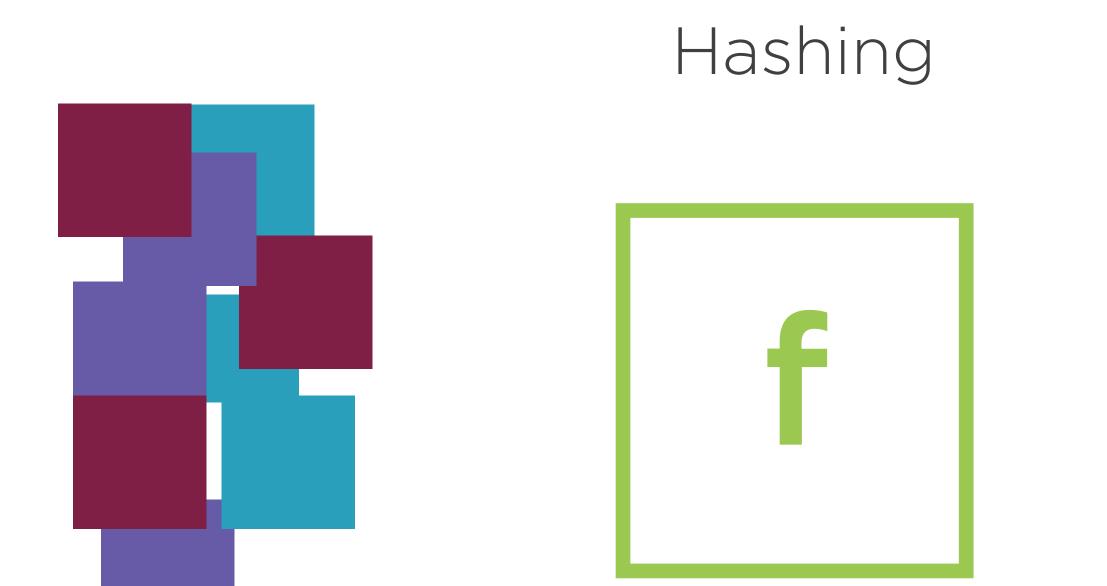
A technique that allows you to lookup specific values very quickly

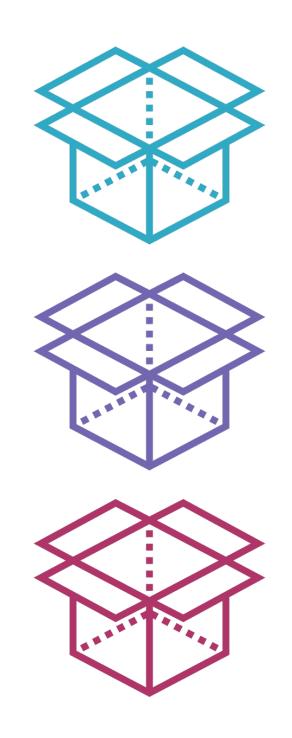


Also can be used to perform dimensionality reduction

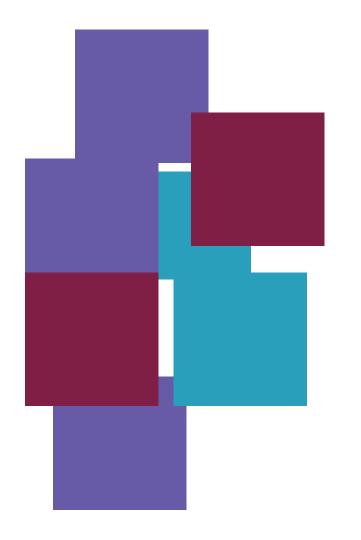


Have a fixed number of categories or buckets

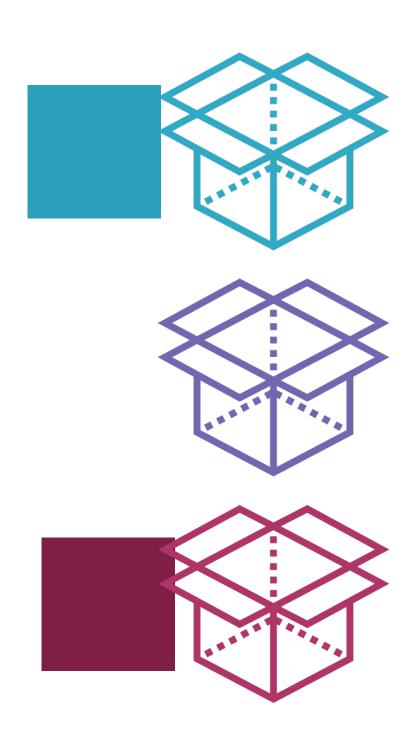


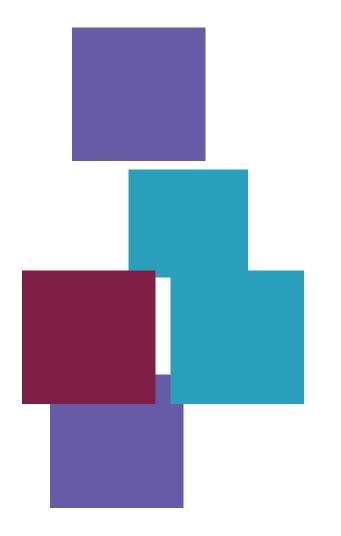


A hash function determines which bucket each value belongs to

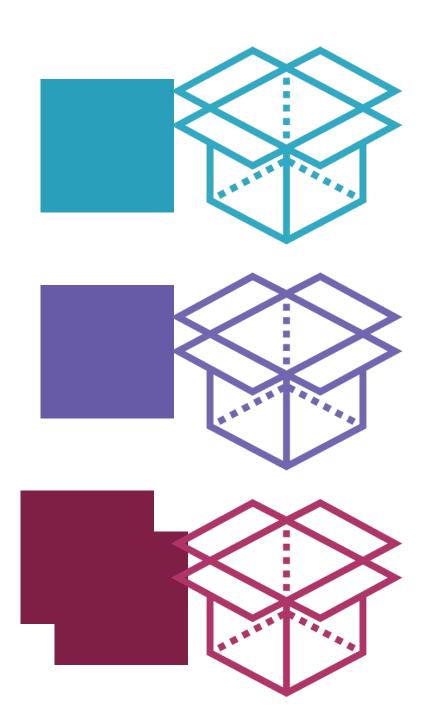






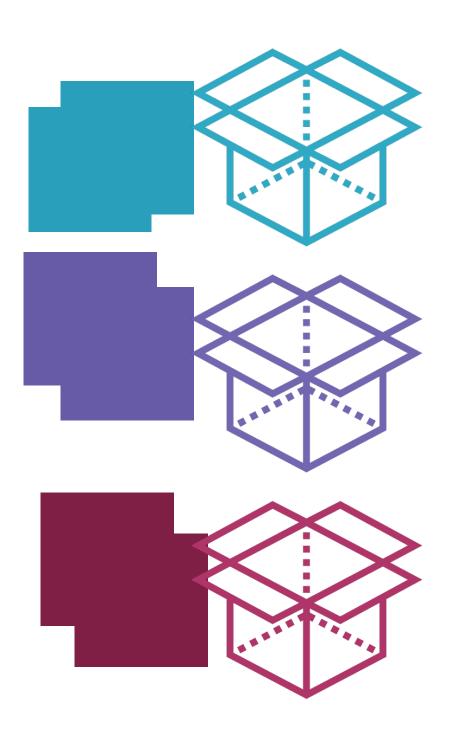




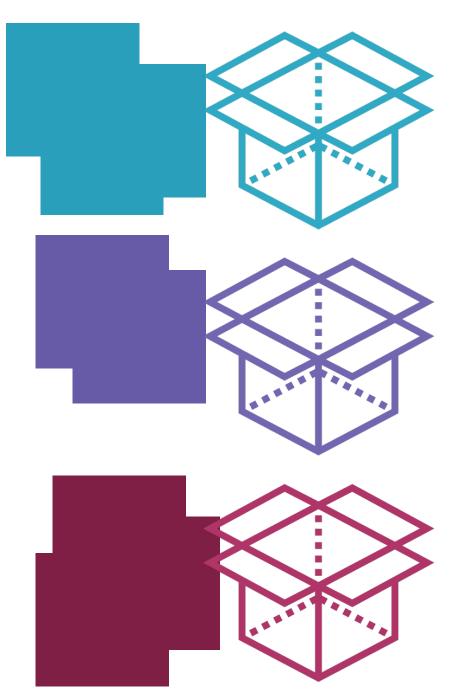


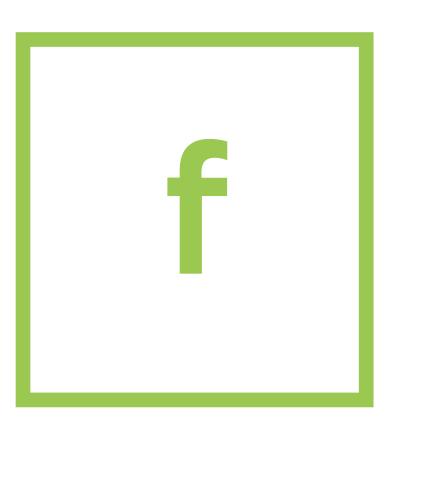


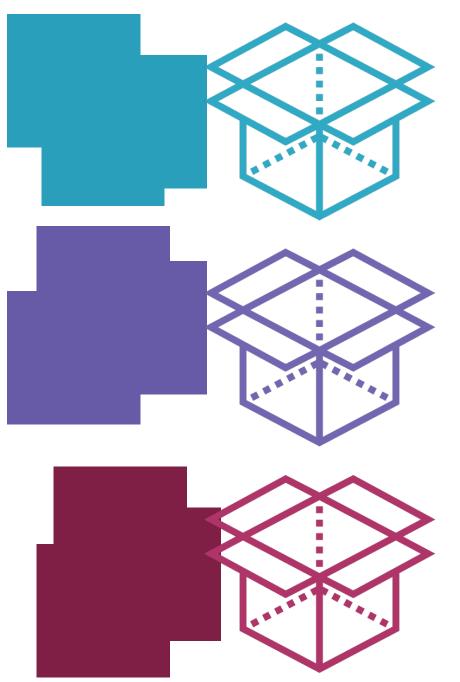


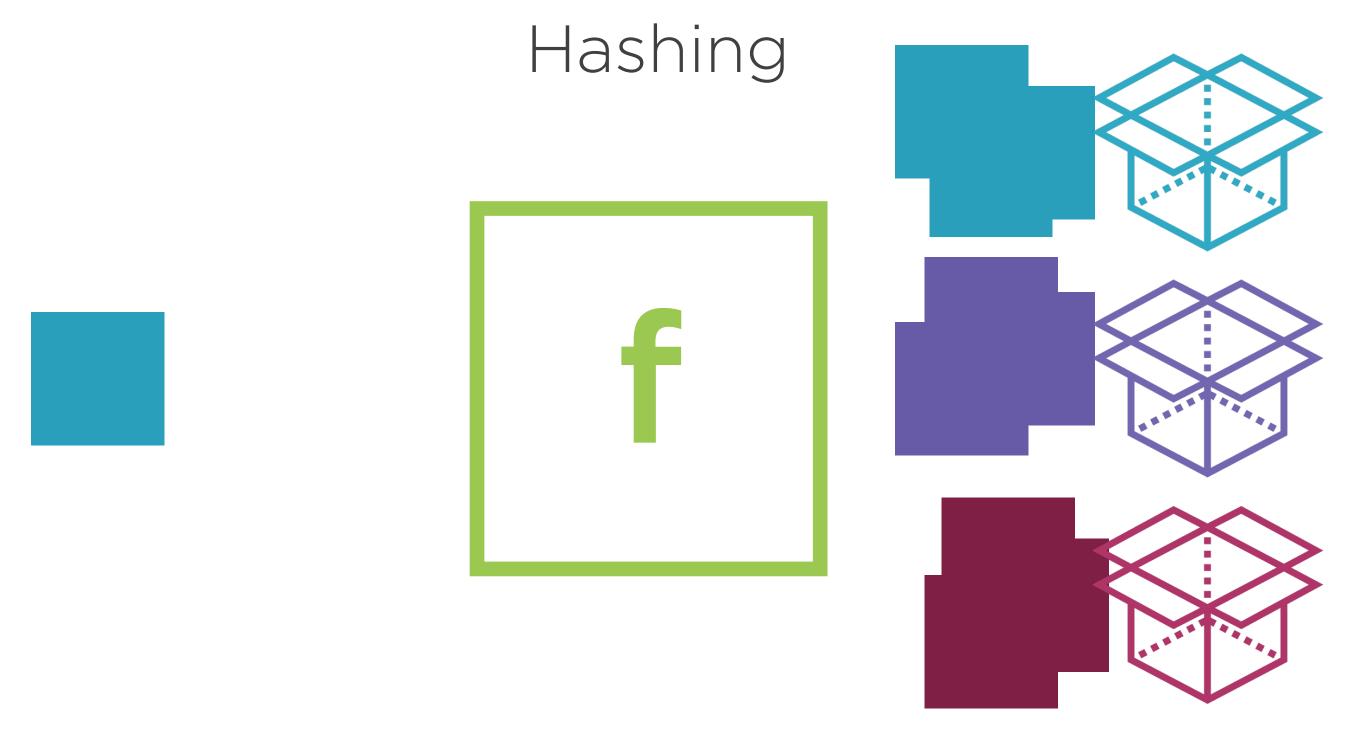




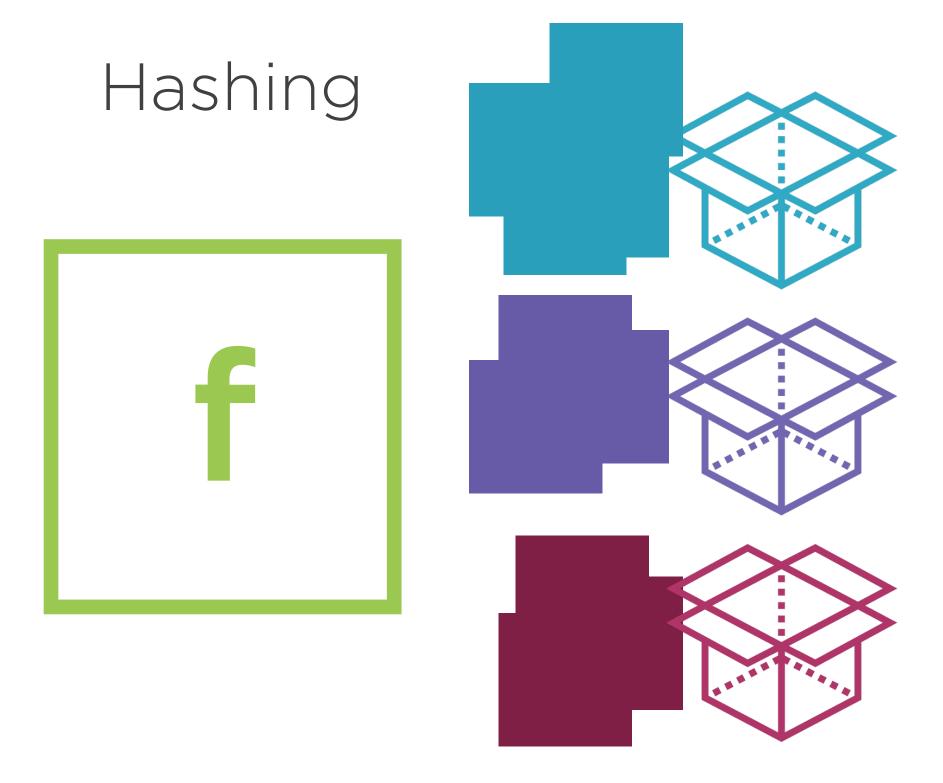




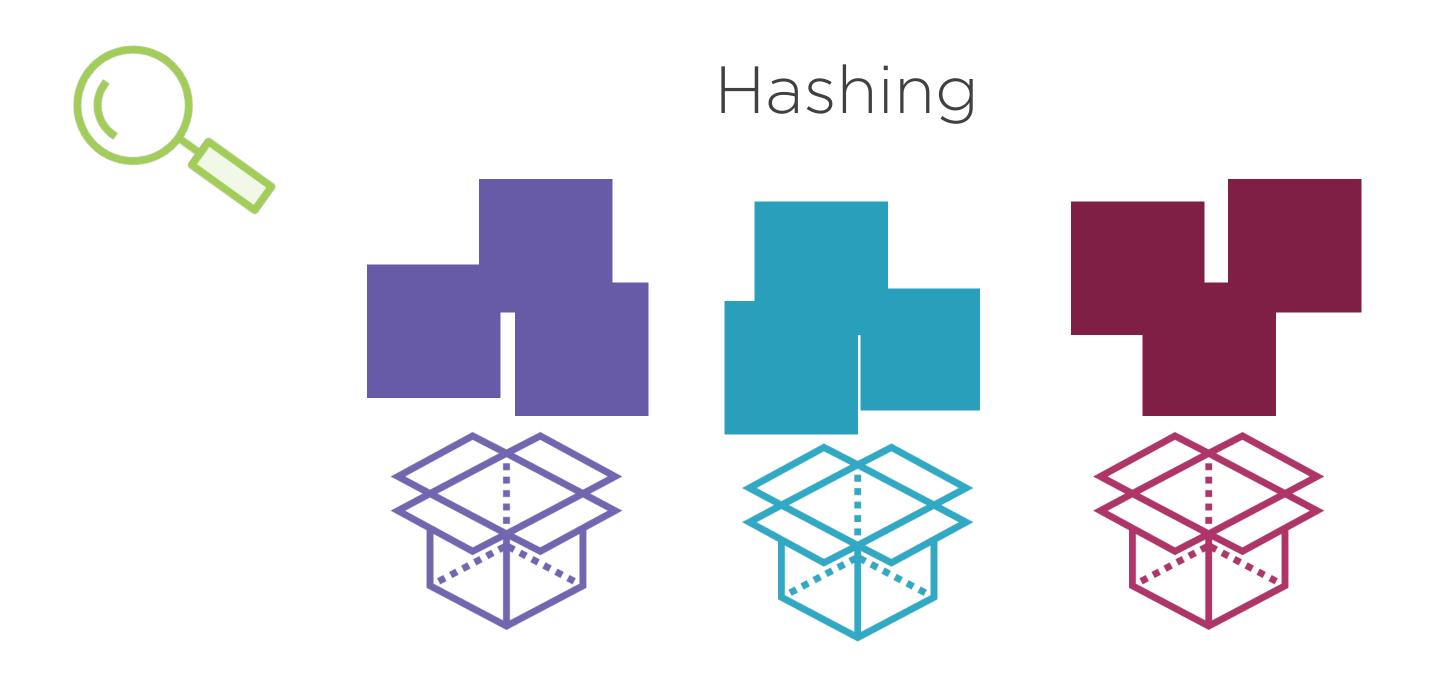




For any new value we know immediately which bucket it belongs to

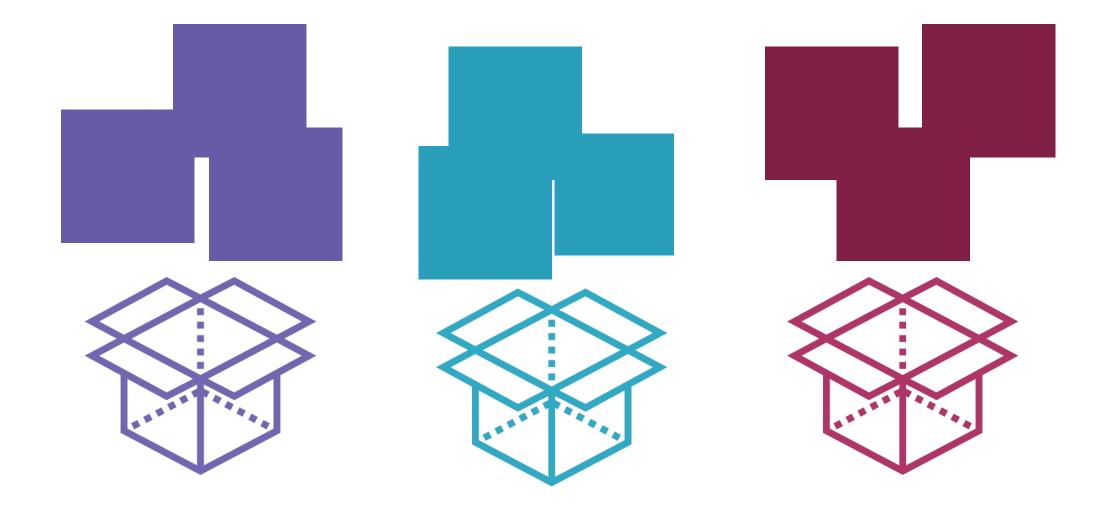


For any new value we know immediately which bucket it belongs to



Each value is hashed so it falls in one of these buckets





A value can only belong to one bucket and always belongs to the same bucket

# Feature Hashing in Text

Apply a hash function to words to determine their location in the feature vector representing a document. Fast and memory efficient but has no inverse transform.

# Dimensionality Reduction



Input: N-dimensional data

Output: k-dimensional data

Where k < N



Input: N-dimensional data

**Output: 1-dimensional data** 

Output is the hash bucket the data maps to



Input: N-dimensional data

Output: k-dimensional data

Can easily extend hashing to output desired dimensionality

## Demo

Converting nominal data to numeric form using feature hashing

# Summary

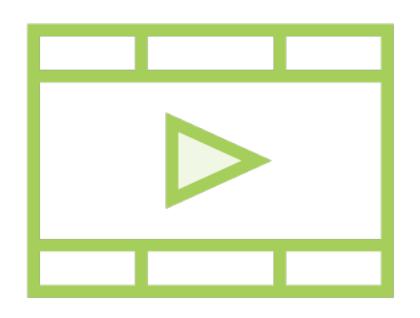
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#### Related Courses



Building Features from Numeric Data
Building Features from Image Data
Building Features from Text Data