

# Tutorial 9 - Class Imbalance & K-means Clustering

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Monday 22<sup>nd</sup> November, 2021

# Disclaimer: Recorded Tutorials will be Posted

## Privacy Preservation:

- Ask questions in the chat<sup>1</sup>
- Keep video off

**Note:** If the above *hinders your ability to learn*  $\wedge$  *violates your privacy*, please let me/Dr. Green know ASAP and video will be post-processed accordingly.

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<sup>1</sup>I encourage unmuted/voice-based questions at any time, but know that this isn't explicitly privacy-preserving

# ML Weekly

Recent news events from the ML community

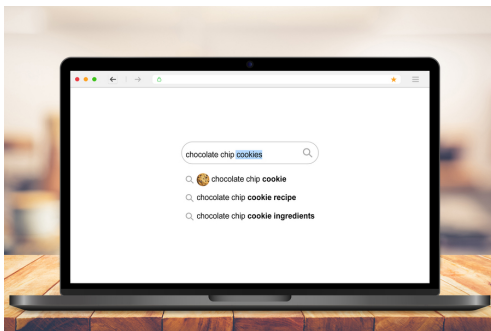
# ML Weekly

1. **(ML)** “Deepfaking the Mind” Could Improve Brain-Computer Interfaces for People with Disabilities



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2. **(AI)** Artificial intelligence sheds light on how the brain processes language



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3. **(AI)** Giving robots social skills



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1. **(ML)** “Deepfaking the Mind” Could Improve Brain-Computer Interfaces for People with Disabilities
2. **(AI)** Artificial intelligence sheds light on how the brain processes language
3. **(AI)** Giving robots social skills
4. **(NLP)** Toward speech recognition for uncommon spoken languages



# Tutorial Overview

We will cover two main concepts in the notebook today:

1. Class Imbalance and K-means clustering



# Tutorial Intuition

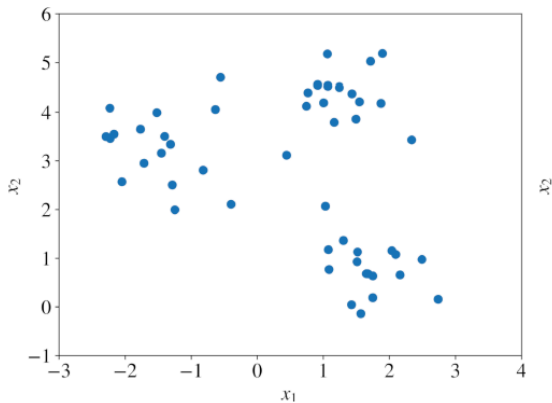
Building an Intuition for the Concepts of this Tutorial

# Animation: Variation in Class Imbalance

# K-means Clustering Intuition

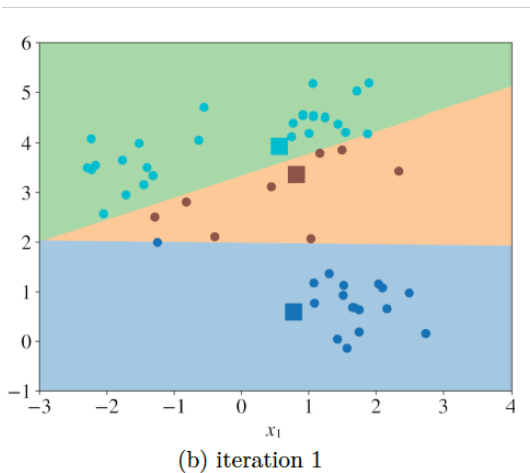
1. Choose  $k$  (the number of clusters)
2. Randomly place  $k$  feature vectors, called centroids, to the feature space
3. Compute the distance from each example  $x$  to each centroid  $c$  using some metric, (ex. Euclidean distance)
4. Then we assign the closest centroid to each example
5. For each centroid, we calculate the average feature vector of the examples labeled with it. The average feature vectors become the new locations of the centroids.
6. Repeat 3-5 until the assignments don't change after the centroid locations were recomputed

# K-means Clustering Intuition

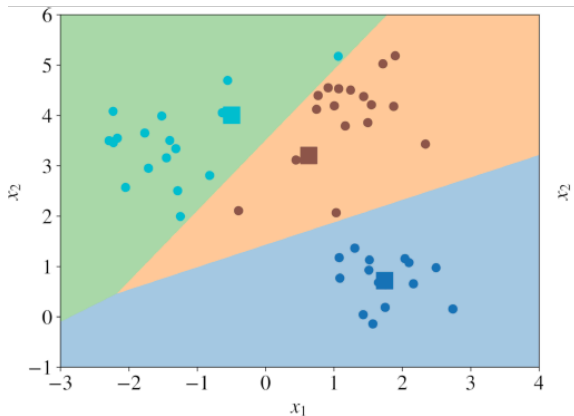


(a) original data

# K-means Clustering Intuition

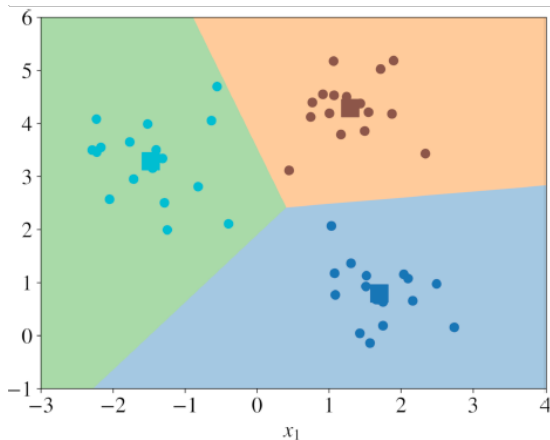


# K-means Clustering Intuition



(c) iteration 3

# K-means Clustering Intuition



(d) iteration 5

# Into the Notebooks we Go...

We will cover two notebooks today!

1. Tutorial 9 - Class Imbalance and K-means clustering



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