



#### Galaxy Morphology Classification using Unsupervised Machine Learning

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#### Overview

- Motivation
- Background
- Methodology
- Results
- Future Work & Challenges

## Motivation

- Understanding of Galaxy Evolution
  - Morphology Classes provides insight into Galaxy Structure, formation & evolution
- Overabundance of data from new state-of-the-art telescopes like James Webb Space Telescope (JWST)
  - Classification of Galaxy was previously conducted by only a handful of expert astronomers
  - Reliance on Citizen Science Projects like Galaxy Zoo for classification

Objective: Classify Galaxies in JWST Images using Unsupervised Machine Learning Method

## Background

- Galaxy morphological classification
  - Categorize galaxies into groups based on their visual appearance
  - The Hubble Sequence invented by Edwin Hubble in 1936, is a morphological classification scheme for galaxies



#### Hubble Tuning Fork Diagram

Ref: https://www.conceptdraw.com/examples/fork

# Background



Ref: https://lawtomated.com/supervised-vs-unsupervised-learning-which-is-better/



Ref: https://towardsdatascience.com/machine-learning-types-2-c1291d4f04b1



# Background

- Centroid-Based Clustering
  - K-Means
- Density-Based Clustering
  - DBSCAN (Density-Based Spatial Clustering of applications with noise)
- Distribution-Based Clustering
  - Gaussian Mixture Models
- Hierarchical-Based Clustering
  - Agglomerative Hierarchical Clustering

#### Centroid-Based Clustering (K-Means)



### **Density-Based Clustering**



#### Distribution-Based Clustering



### Hierarchical-Based Clustering



Agglomerative Hierarchical Clustering

## Overview of Clustering Results



Ref: https://scikit-learn.org/stable/modules/clustering.html

## Methodology

- Simulated JWST mosaics (large surveys of the sky) are used to create stamps (individual cutouts of galaxies)
  - IllustrisTNG100 with added noise to simulate images taken from JWST
  - Correspond to those taken via JWST's NIRCam between wavelength range of 1-4 microns
    Filters: F115W, F150W, F200W, F277W, F356W, F444W
- A total of 10602 stamps were generated (1767 stamps created in each Filter)
  - Format: FITS (Flexible Image Transport System)
  - Image Size: 64 x 64 pixels
- Vector Quantized Variable Autoencoder (VQ-VAE) is used to reduce the dimensionality of the stamp data into a discrete latent representation
- Clustering of the lower dimension image data is done using a clustering algorithm.
  - K-means •

# Results

Original



#### Latent Space



Reconstruction



## Results



#### Future Work & Challenges

- Because JWST is still in its infancy, there is very little labeled morphology data to validate prediction/classification accuracy
  - Other methods to evaluate clustering results currently being explored
- Reliability of K-means on this dataset
  - More research into K-means should be explored to understand impact on highdimensional data
  - Testing results from other types of algorithms might provide different insight
  - Try different clustering algorithm/methodology that might provide different results i.e Hierarchical Clustering, HDBSCAN, Spectral Clustering, etc.

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#### Questions?

## Quiz Time!

• Go to mycourses to answer the question