



# Capstone Project - Plant Recommendation Engine

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# Background



- People are increasingly turning to gardening as a hobby
  - [Note record sales of seeds in 2020](#)
- Starting a new hobby can be daunting
- Lots of questions about what plants to grow
  - More difficult to answer than it seems at first
- Lots of information online about growing plants

**Problem Statement:**

**Create a model that can recommend  
plant species to grow based on  
growing conditions**

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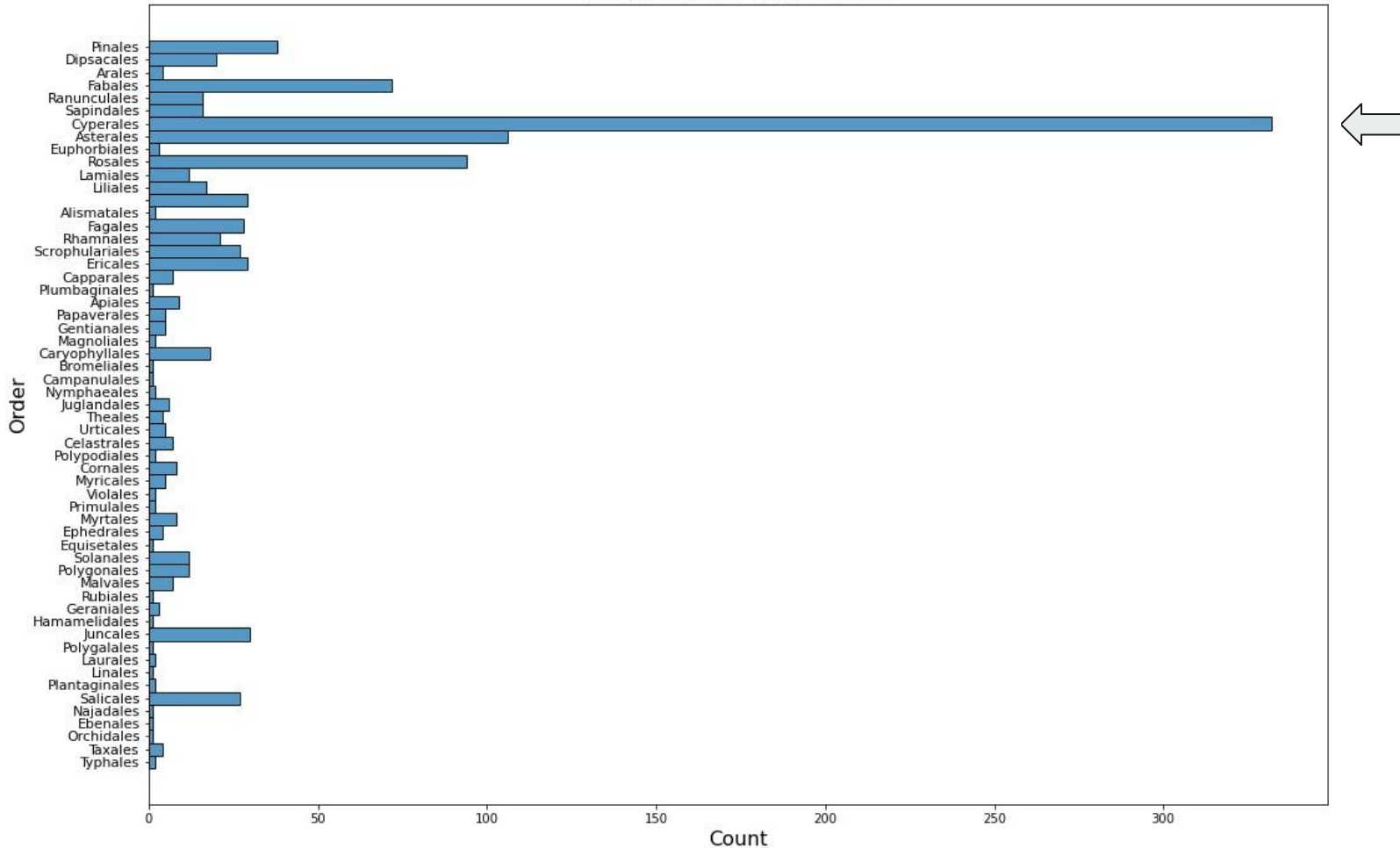
# Data Collection

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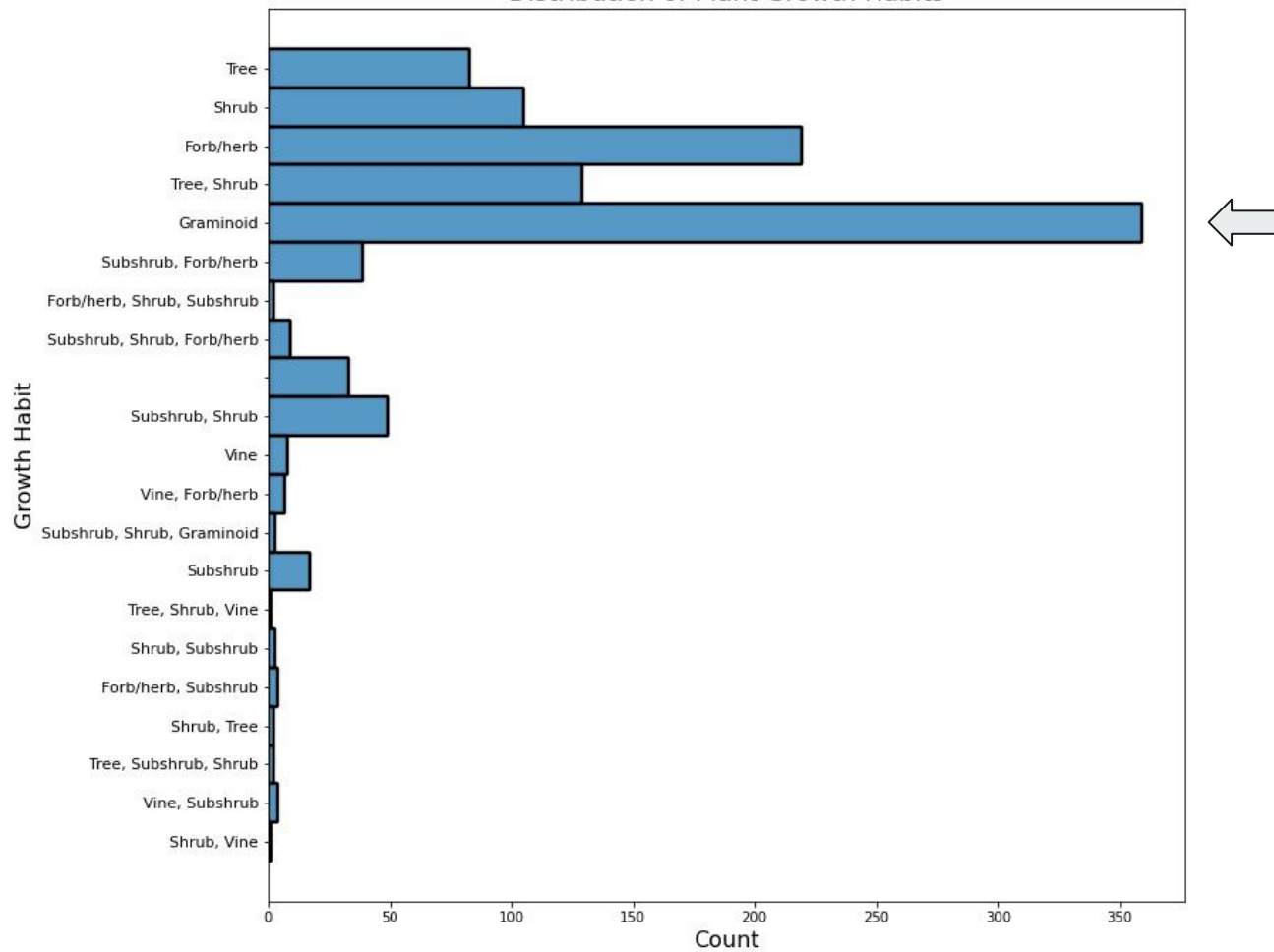
- Surprisingly difficult to find a good database
  - Many paywalls and databases lacking detail
- Used the USDA's PLANTS database
- Information about ~90,000 species across North America
- Detailed information about **1078 species**
- Accessed this via a SQL query
  - Database came from a USDA API
- Data cleaning
  - Removed null values
  - One-Hot Encoding categorical features



# Distribution of Plant Orders



# Distribution of Plant Growth Habits



# Features Included in the Model



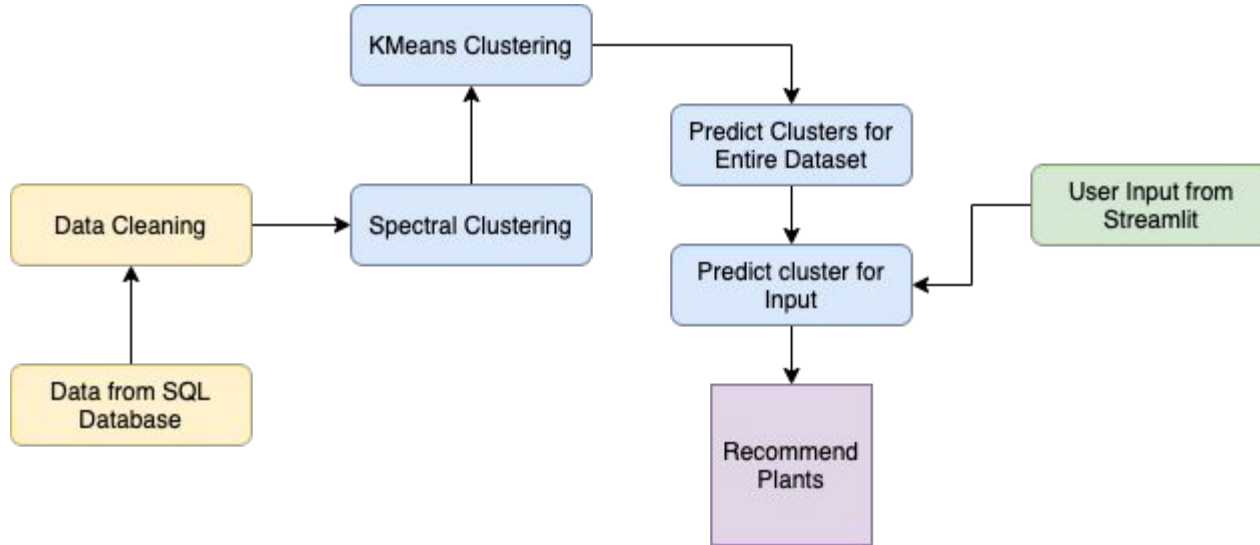
- Growth Habit
- Family
- Native Status - whether or not a plant is native to a region
- Toxicity
- Shade Tolerance - how much sun does a plant need?
- Minimum Temperature (F) - The coldest temperature a species can survive
- Soil pH
- Drought Tolerance
- **22 features in total!**



# Modeling

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# An Overview of the Modeling Process



# Clustering



- Tried out several clustering models
- DBSCAN
  - Helped figure out the number of clusters needed (recommended 68 clusters)
  - Silhouette Score of 0.14
- KMeans
  - Silhouette Score of 0.049
- Went with Spectral Clustering
  - Creates a graphical representation of the dataset
  - Works with KMeans to predict clusters
  - Outperforms other clustering models regularly, but is more error prone
  - **Silhouette Score of 0.51!**

# Building the Recommendation Engine



1. Make a dummy entry based on the user input
  2. Use clustering model to predict cluster locations for the whole dataset
  3. Predict the cluster of the dummy entry
  4. Sample plants randomly from the same cluster as the dummy entry
  5. Display the species of the plants suggested
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- Item-based Recommender System
  - Created a test case purely in Python first

# Creating an App



- Lots of different options available here
- Decided to go with **Streamlit**
  - New library/ hosting service
  - Can be used to turn Python scripts into interactive apps
- My Streamlit app can be found on this project's repo
  - If you want to run the app, make sure to install Streamlit via Conda/pip and download the entire project repo

# Conclusions

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# Results



- Successfully created a plant species recommendation engine
- Very bare bones recommender, however
- Few distinct clusters make the results sometimes inaccurate
- App runs slowly
- Recommender would be useful for:
  - Large commercial clients
  - Native plant restoration

# Next Steps and Recommendations



- Try to get more data on more commercially available plants
  - Too many grasses in this dataset!
- Improve upon the Streamlit app
  - Add Wikipedia links
  - Add Images for each suggested plant
  - Add more features from the dataset
- Try out a hybrid approach with collaborative filtering
  - Try to include what plants sell in conjunction with each other
  - These models tend to outperform item-based recommenders
- Make the app available to the public



**Thank You!**