



CIS325 FINAL PROJECT: NETFLIX RECOMMENDATION ENGINE

The goal of this project is to simulate an enterprise level data science challenge by engaging a small team of students to conduct an end-to-end process on a chosen dataset. This includes building and deploying a machine learning model as a web service which can be demonstrated as a client-side application.

The selected dataset was sourced from Kaggle¹ and based on publicly available Netflix data from 2021, containing information on 8800+ steaming movies and TV shows at the time. This robust data features details such as movie title, genre, director, cast, release year, country of origin and description of content.

Our model will allow a user to input a movie description of what might be available on Netflix and receive a recommendation for interesting titles, much like what a consumer would experience when navigating the platform on their device.

¹ Online Kaggle source: <https://www.kaggle.com/datasets/shivamb/netflix-shows>

GETTING STARTED

DEPENDENCIES:

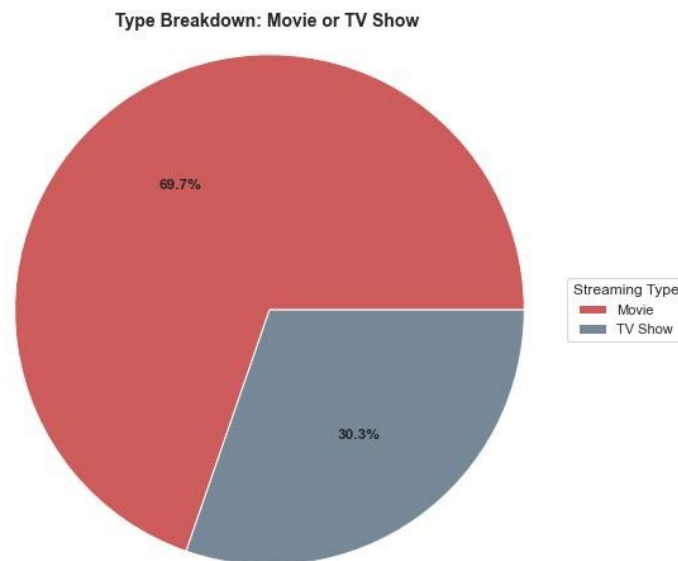
- ❖ Python 3
- ❖ Scikit-Learn
- ❖ Pandas/Numpy
- ❖ Seaborn/Matplotlib
- ❖ Current Web Browser
- ❖ Jupyter Notebook
- ❖ MLflow
- ❖ Tensorflow
- ❖ Pickle
- ❖ Microsoft Azure Portal
- ❖ Flask, Flasgger APIs
- ❖ Docker/CLI

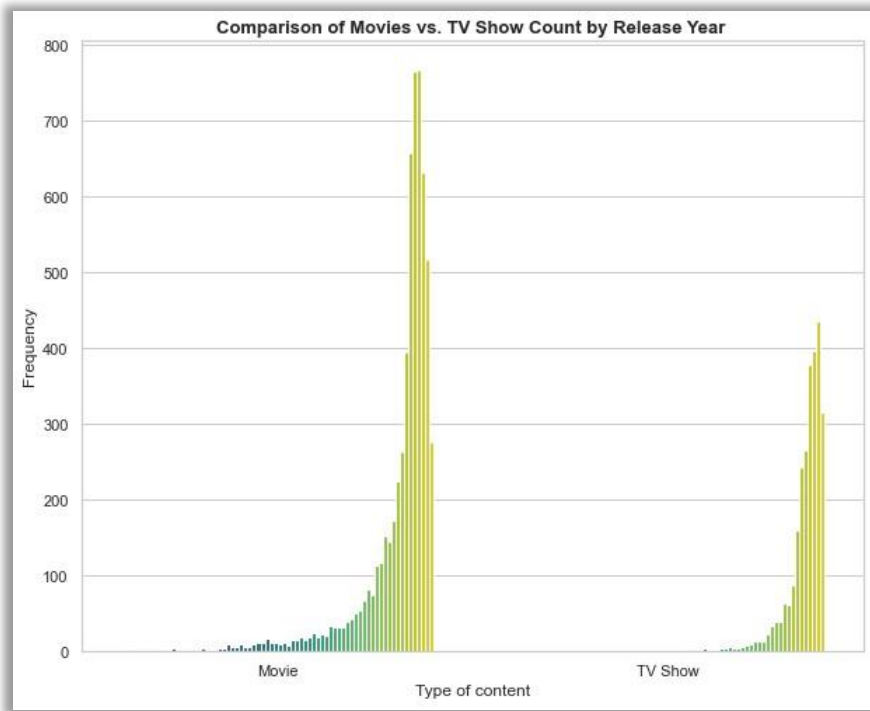
INSTALLATION RESOURCES:

- ❖ Install all relevant Python packages: www.anaconda.com/products/individual (such as Scikit-Learn)
- ❖ Pandas: https://pandas.pydata.org/pandas-docs/stable/getting_started/install.html
- ❖ Install MLFlow package: www.mlflow.org/docs/latest/quickstart.html#installing-mlflow
- ❖ Docker overview: <https://docs.docker.com/desktop/>
- ❖ Container registry setup: <https://docs.microsoft.com/en-us/azure/container-registry/container-registry-get-started-portal#create-a-container-registry>
- ❖ Azure Kubernetes Service: <https://azure.microsoft.com/en-us/overview/kubernetes-getting-started/>

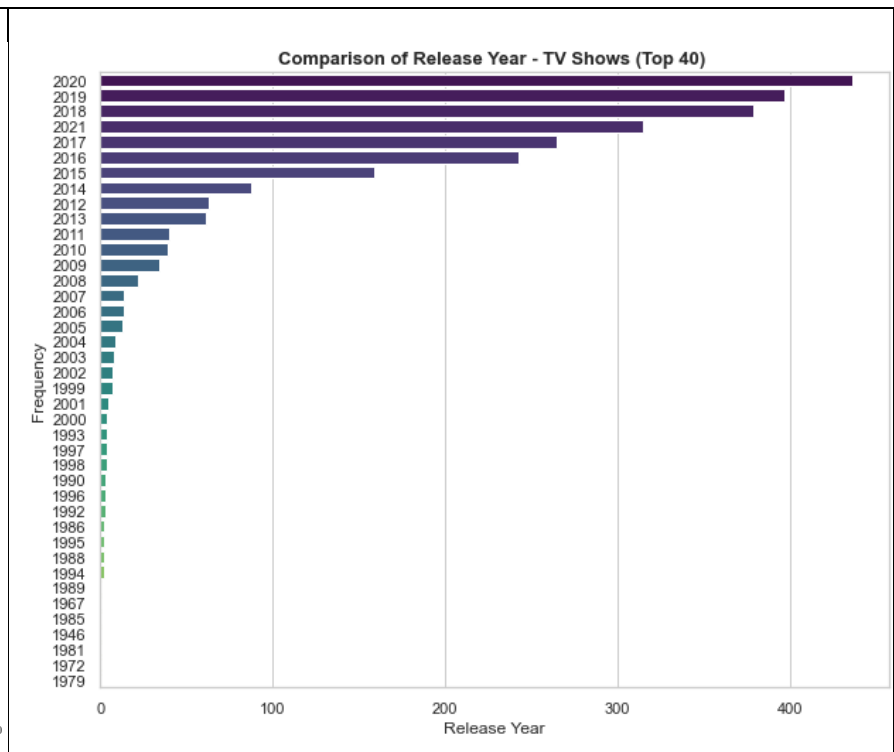
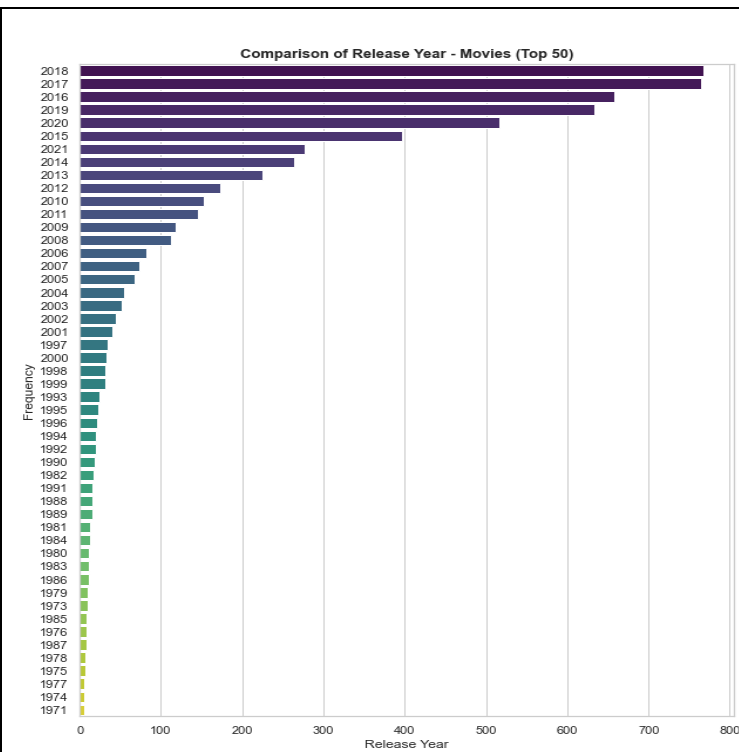
EXPLORATORY DATA ANALYSIS

It is important to note that the bulk of the data pertains to movie content. This could mean that any recommendation engine could offer better accuracy when searching for similar movies since there is a greater pool of data for that type.

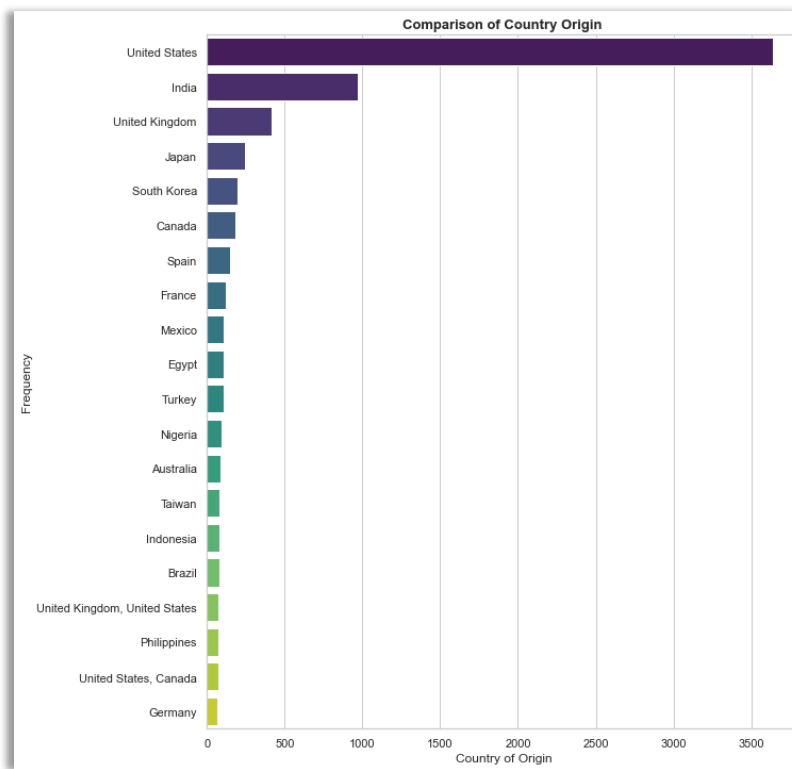




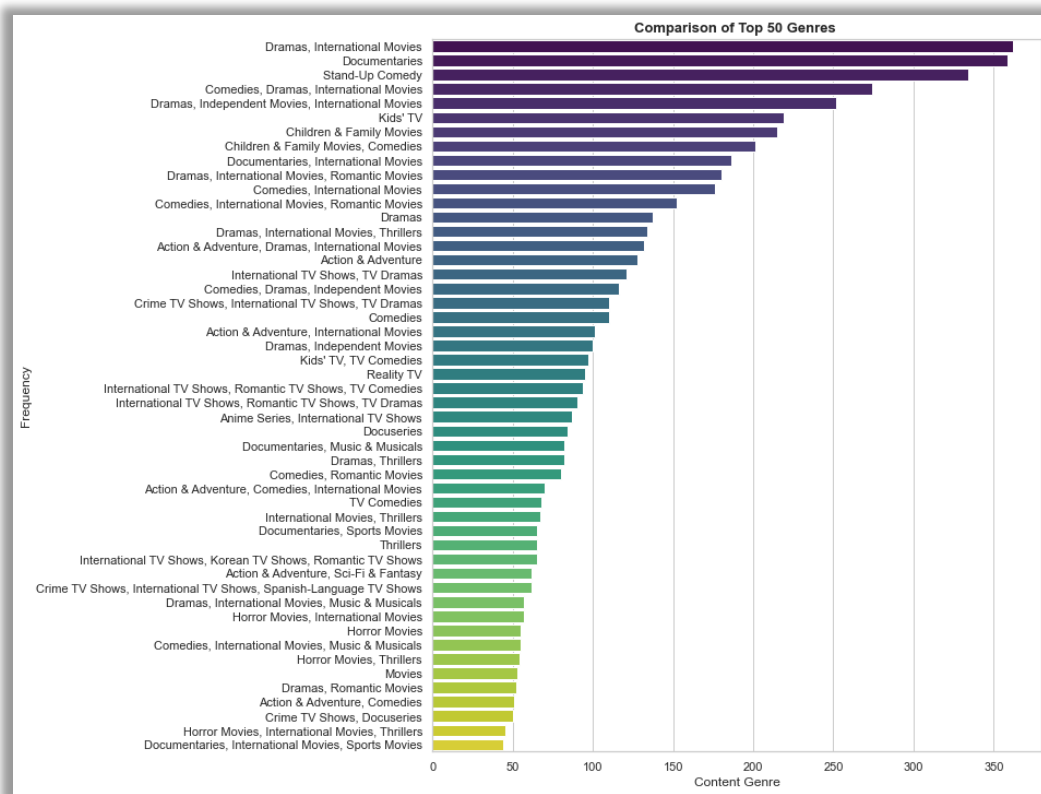
There is definitely an increase in streaming content as time passes except for a sharp decline in the past two years due to the pandemic, however the general curve in the frequency count increase seems very similar between movies and TV shows, even though movies account for far more content overall. There was slightly less of a drop off in TV shows however, as they are easier to produce with smaller production needs.



There is a bit of a discrepancy as more TV shows were released from 2018-2020 than 2021, likely because the data for 2021 is incomplete. More movies were released from 2016-2020 than 2021, also likely because the data for 2021 is not complete. However clearly less movies were produced in 2020 than prior years-- no doubt due to the global pandemic and filming shut downs.

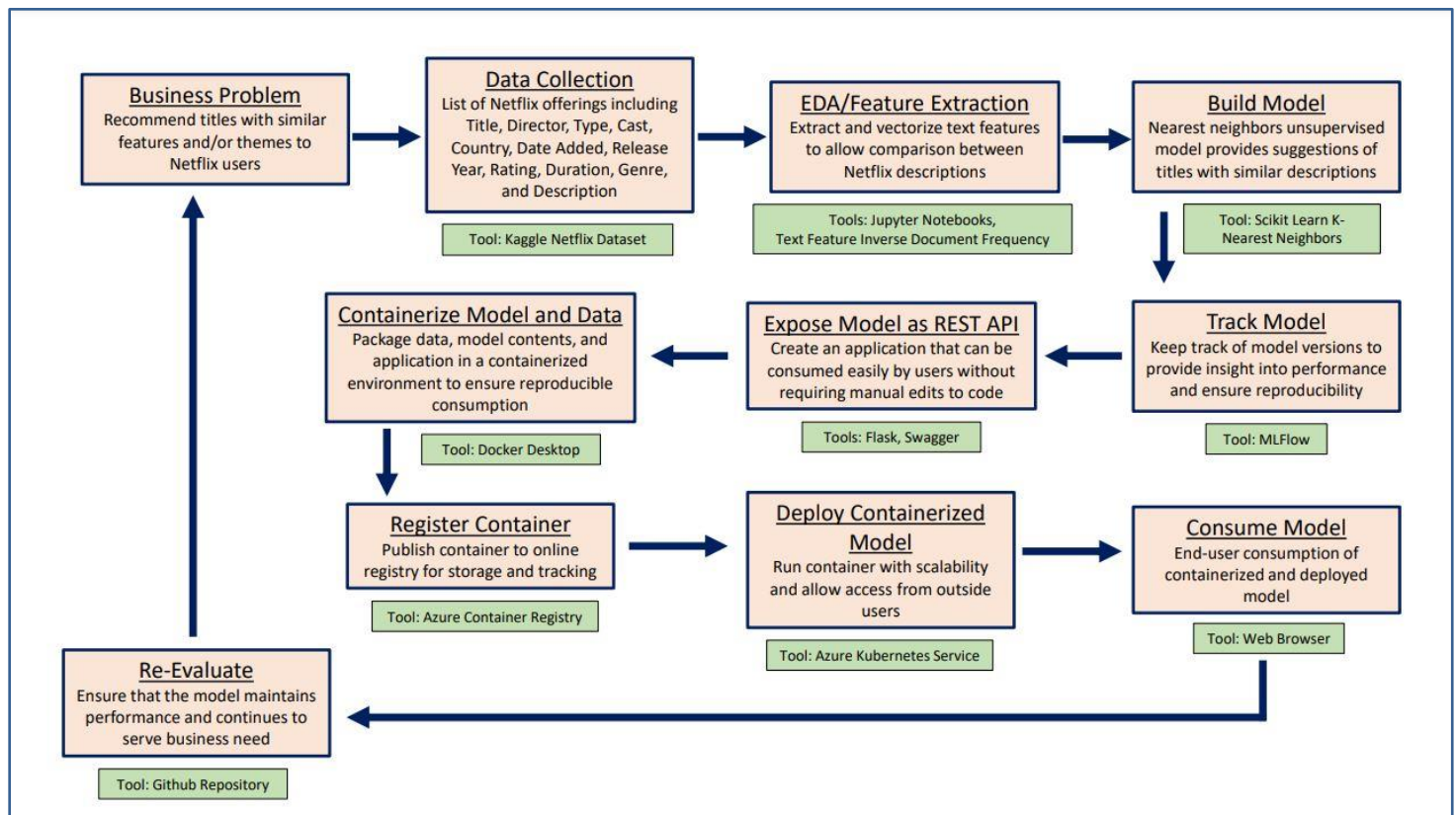


As expected, the US, UK and India (Bollywood) are the top 3 content producers. Countries with well-funded arts programs are also in the top 10 (Canada, France, Japan, South Korea) however I am surprised that Nigeria is not higher up in the list since they have Nollywood- but that just means Netflix is not releasing much of their content. It's also surprising that Spanish-speaking countries are not better represented since there are a lot of Spanish-speaking consumers for streaming content.



Clearly dramas of all kinds, comedies, romance, documentaries and international films are highly represented. These categories may perform best in the recommendation engine since the data pool for them will be large.

PROJECT DESIGN FLOWCHART

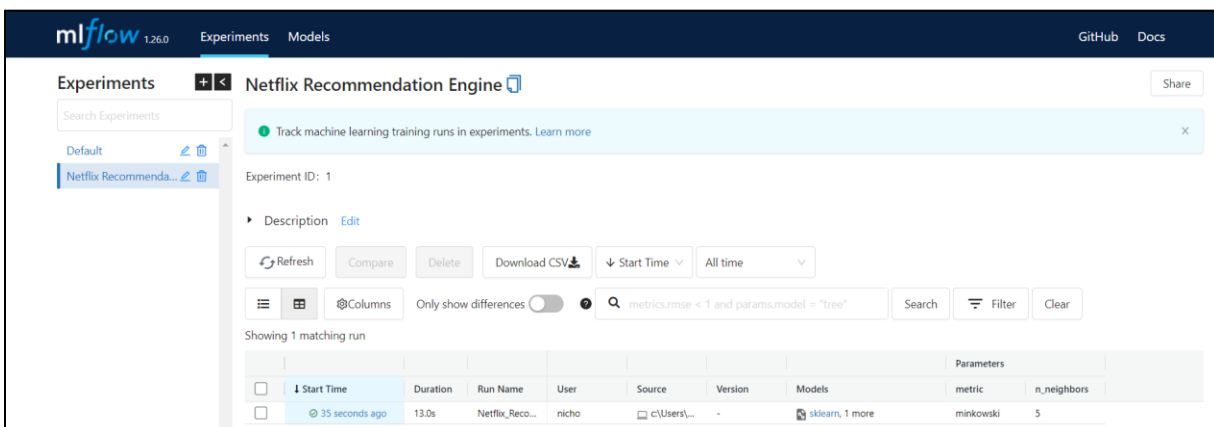


MODEL CREATION

As a first step, the Description column containing the text of the movie synopsis was vectorized and converted into a sparse matrix which contains the movie titles on the index and word or phrase vectors as column features. While cosine similarity and K-means clustering were explored as options to create the recommendation engine, ultimately a K-Nearest Neighbors model was used to expose a model via REST API. A model creation function was employed to search the matrix for the 5 'nearest neighbors' and provide similar movie or show titles to the end user.

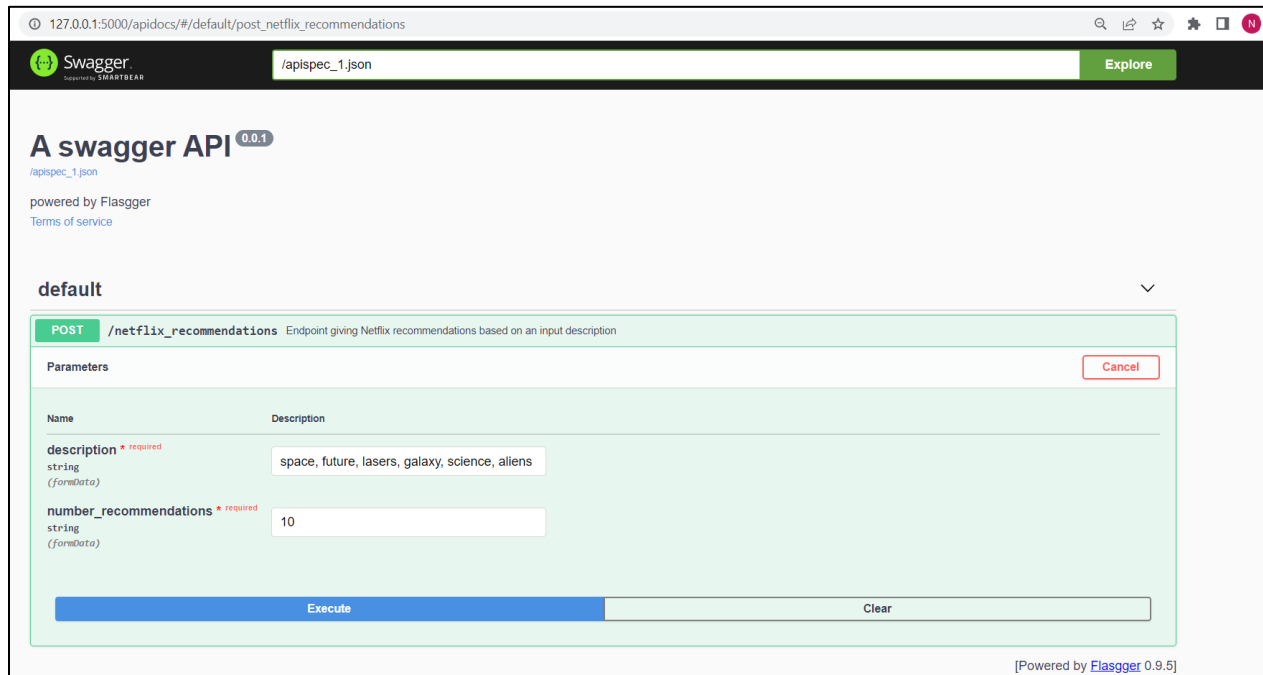
MLFLOW DASHBOARD

Once a new experiment has been created and MLFlow tracking has been established, the MLFlow portal will list any model runs (successful or not) as seen below in the experiment tracking interface.



SWAGGER/REST API TESTING

Once a model is successfully tracked and saved via MLFlow, it should be tested locally by creating a Flask python file (.py) and exposed on a local port (such as 5000) using the Swagger API, entering the proper input to test the expected output.



In this case, the Flask app file was programmed to print output (the requested amount of recommended titles) to the command line interface. Three different sets of inputs were tested and the movie recommendations can be seen printed to the CLI below.

```
Anaconda Prompt (anaconda3) - python Netflix_Recommender_Flask_v1_NSH_05.24.22.py
(base) C:\Users\nicho>cd C:\Users\nicho\CIS325\Final Project\final_project_app
(base) C:\Users\nicho\CIS325\Final Project\final_project_app>python Netflix_Recommender_Flask_v1_NSH_05.24.22.py
* Serving Flask app "Netflix_Recommender_Flask_v1_NSH_05.24.22" (lazy loading)
* Environment: production
  WARNING: This is a development server. Do not use it in a production deployment.
  Use a production WSGI server instead.
* Debug mode: on
* Restarting with watchdog (windowsapi)
* Debugger is active!
* Debugger PIN: 106-552-354
* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
127.0.0.1 - - [26/May/2022 18:42:20] "GET /apidocs/ HTTP/1.1" 200 -
127.0.0.1 - - [26/May/2022 18:42:20] "GET /apispec_1.json HTTP/1.1" 200 -
{0: 'A StoryBots Space Adventure', 1: 'The Mars Generation', 2: 'Star Trek: Deep Space Nine', 3: 'Antariksha Ke Rakhwale', 4: 'Pocoyo Halloween: Space Halloween', 5: 'The Epic Tales of Captain Underpants in Space', 6: 'Skylines', 7: 'Chhota Bheem: Bheem vs Aliens', 8: 'Lockout', 9: 'Rocko's Modern Life: Static Cling'}
127.0.0.1 - - [26/May/2022 18:42:43] "POST /netflix_recommendations HTTP/1.1" 200 -
{0: 'Theeya Velai Seyyanum Kumaru', 1: 'Find Yourself', 2: 'Kajraare', 3: 'Life Plan A and B', 4: 'Iyore'}
127.0.0.1 - - [26/May/2022 18:43:39] "POST /netflix_recommendations HTTP/1.1" 200 -
{0: 'Cells at Work!', 1: 'Mad Ron's Prevues from Hell', 2: 'Hell and Back', 3: 'O-Negative, Love Can't Be Designed', 4: 'Post Mortem: No One Dies in Skarnes', 5: 'The Babysitter: Killer Queen', 6: 'Primal Fear'}
127.0.0.1 - - [26/May/2022 18:44:34] "POST /netflix_recommendations HTTP/1.1" 200 -
```


CONTAINERIZATION WITH DOCKER

The model and all of the relevant data and dependencies are then containerized using Docker.

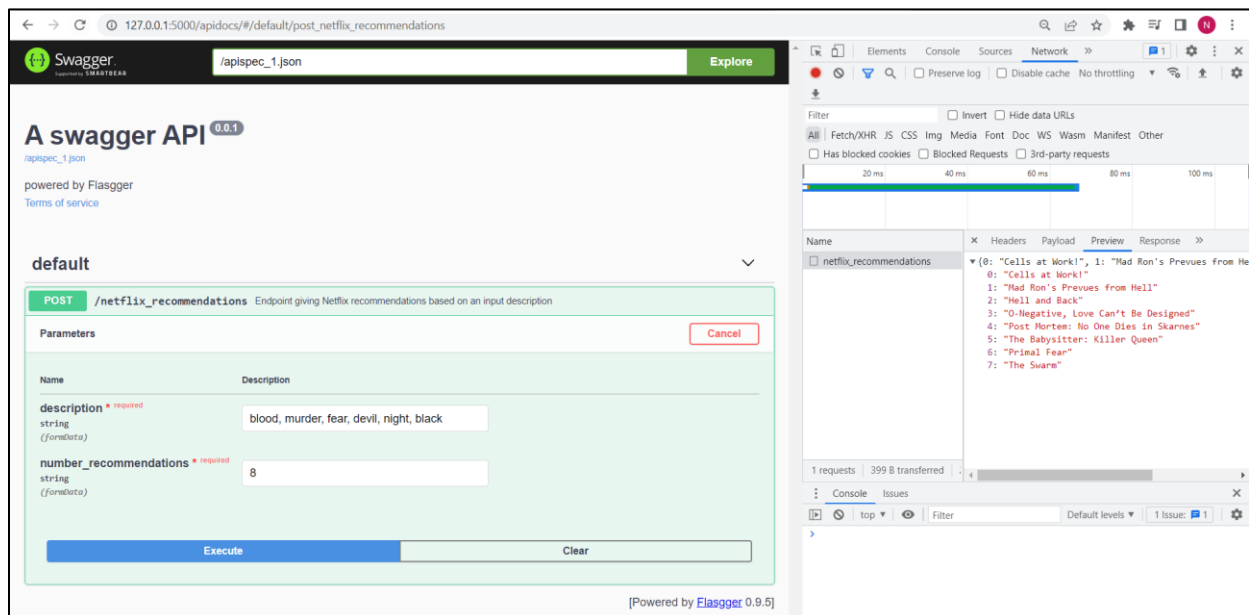
```
Command Prompt - docker run -p 5000:5000 netflix-recommender-api

C:\Users\nicho\CIS325\Final Project>docker build -t netflix-recommender-api .
[+] Building 54.9s (9/9) FINISHED
=> [internal] load build definition from Dockerfile                                0.0s
=> => transferring dockerfile: 32B                                              0.0s
=> [internal] load .dockerignore                                                 0.0s
=> => transferring context: 2B                                                  0.0s
=> [internal] load metadata for docker.io/library/python:3.8.8                 2.5s
=> [internal] load build context                                                0.0s
=> => transferring context: 3.16kB                                             0.0s
=> CACHED [1/4] FROM docker.io/library/python:3.8.8@sha256:e84c219fe873ab169551469f32b57facf7d7bade941ccf0cbcc5 0.0s
=> [2/4] COPY ./final_project_app /usr/local/python/                          0.0s
=> [3/4] WORKDIR /usr/local/python                                             0.0s
=> [4/4] RUN pip install -r requirements.txt                                    49.7s
=> exporting to image                                                         2.5s
=> => exporting layers                                                         2.5s
=> => writing image sha256:cf888d8578207bd54cf68383a65dee1597fc92ba643ccaaff58e5435b31f7058 0.0s
=> => naming to docker.io/library/netflix-recommender-api                     0.0s

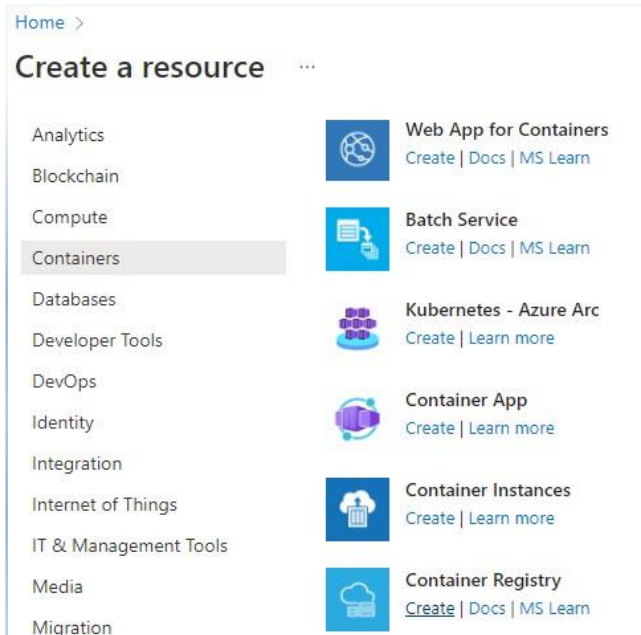
Use 'docker scan' to run Snyk tests against images to find vulnerabilities and learn how to fix them

C:\Users\nicho\CIS325\Final Project>docker run -p 5000:5000 netflix-recommender-api
* Serving Flask app 'Netflix_Recommender_Flask' (lazy loading)
* Environment: production
WARNING: This is a development server. Do not use it in a production deployment.
Use a production WSGI server instead.
* Debug mode: off
* Running on all addresses (0.0.0.0)
WARNING: This is a development server. Do not use it in a production deployment.
* Running on http://127.0.0.1:5000
* Running on http://172.17.0.2:5000 (Press CTRL+C to quit)
172.17.0.1 - - [27/May/2022 01:44:17] "GET /apidocs/ HTTP/1.1" 200 -
172.17.0.1 - - [27/May/2022 01:44:17] "GET /apispec_1.json HTTP/1.1" 200 -
172.17.0.1 - - [27/May/2022 01:44:17] "GET /flasgger_static/favicon-32x32.png HTTP/1.1" 200 -
```

An example of consuming Flask app with a Docker container:



CONTAINER REGISTRY

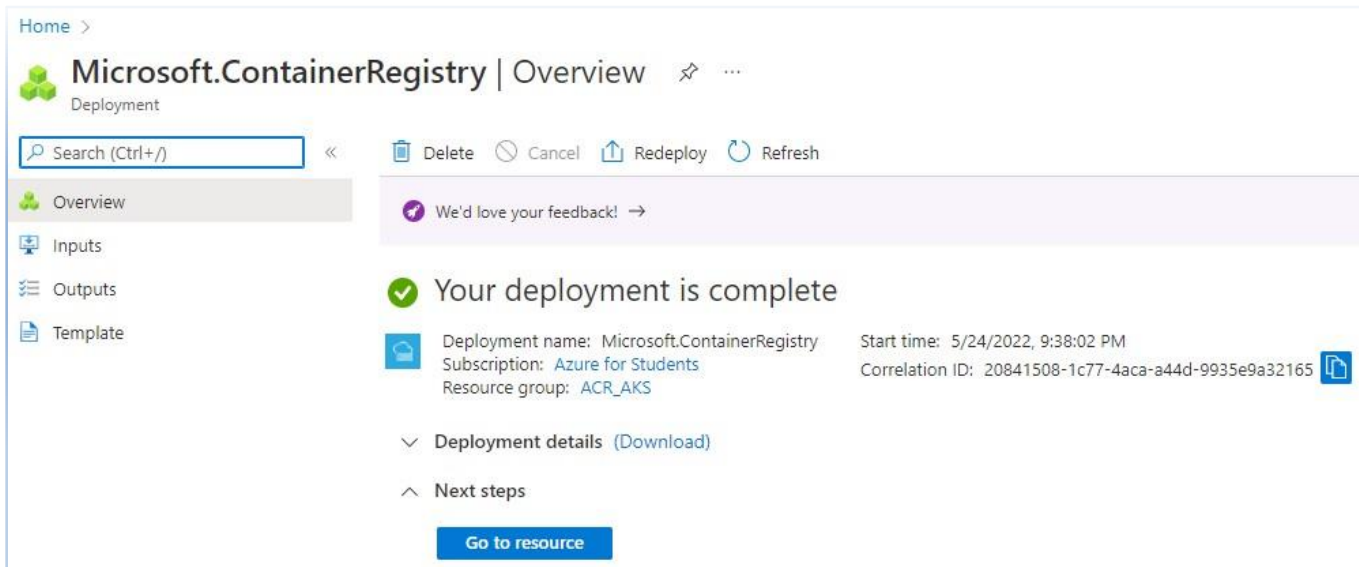


Creating a container registry on ML Azure is as simple as creating a new virtual machine. The first step is to log into the portal and choosing the option to create a new resource, and then searching the options for “Container Registry”.

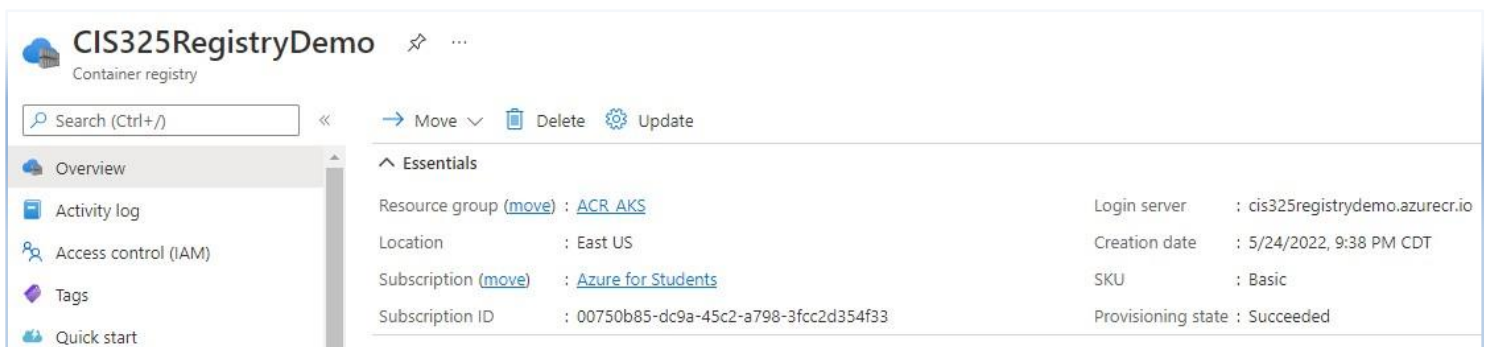
Select ‘create’ to launch the next step.

The first and most important step of creating the registry is to choose the subscription option, name the new resource group, name the registry, choose the region location and a Basic SKU for our purposes. For the rest of the screens, the default options will suffice.

Review and create the resource. The final screen will confirm that deployment is complete.



After the Container Registry has been successfully deployed, on the resource home screen you will be able to view valuable information for future use as seen below:



It can also be viewed on the command line interface by invoking the **az acr list** command as shown below.

```
C:\Users\nicho>az aks get-credentials --resource-group CIS325FinalProject --name cis325akc
Merged "cis325akc" as current context in C:\Users\nicho\.kube\config

C:\Users\nicho>kubectl get nodes
NAME                                STATUS    ROLES    AGE   VERSION
aks-agentpool-51973739-vmss000000 Ready    agent    59m   v1.22.6

C:\Users\nicho>az acr list --resource-group CIS325FinalProject --query "[].{acrLoginServer:loginServer}"
[
  {
    "acrLoginServer": "cis325acr.azurecr.io"
  }
]
```

CONTAINER REGISTRATION WITH KUBERNETES

The following steps show how to build and register an image in the previously created container registry.

First, log into Azure via the CLI and run the command **az login**.

```
C:\Users\nicho>az login
A web browser has been opened at https://login.microsoftonline.com/organizations/oauth2/v2.0/authorize. Please continue
the login in the web browser. If no web browser is available or if the web browser fails to open, use device code flow w
ith 'az login --use-device-code'.
{
  "cloudName": "AzureCloud",
  "homeTenantId": "7b6c0dd8-ebf0-43d4-8182-319c517ae8b8",
  "id": "f4571f50-062f-41b9-9544-667164e98a8e",
  "isDefault": true,
  "managedByTenants": [],
  "name": "Azure for Students",
  "state": "Enabled",
  "tenantId": "7b6c0dd8-ebf0-43d4-8182-319c517ae8b8",
  "user": {
    "name": "nicholasharrison708@outlook.com",
    "type": "user"
  }
}
```

Be sure to have the exact Container Registry name saved, which can be found on the Azure portal, and have the Docker 'daemon' running on your machine.

To authenticate to the Azure ACR, on the command line interface run the following command:

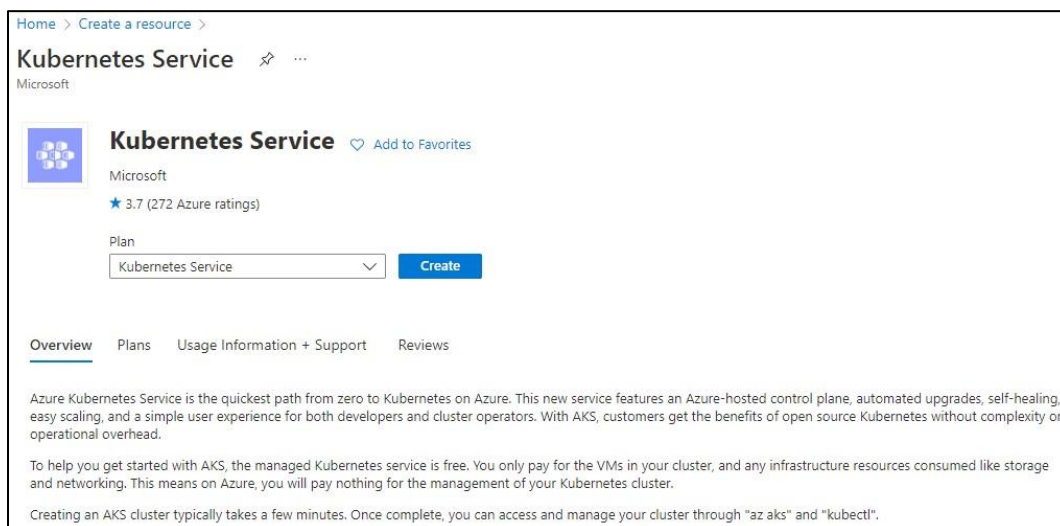
az acr login <registry_name> --name

Be sure you are working in the proper directory where files are stored. If necessary change to where your Dockerfile is located.

Now type the following command to build and register your docker image with the ACR:

az acr build --image myimage:v1 --registry <registry_name> --file Dockerfile .

Now, the Kubernetes cluster must be created. This is similar to creating a VM or the Container registry. Search for a Microsoft Kubernetes Service under available resources and select 'Create'.



It's important to note the proper criteria to enter or select on each screen. On the basics tab, choose the proper resource group with the container registry, use a Standard configuration, name the Cluster, choose the region, available zones and node sizes.

Home > Create a resource > Kubernetes Service >

Create Kubernetes cluster ...

Project details

Select a subscription to manage deployed resources and costs. Use resource groups like folders to organize and manage all your resources.

Subscription * ⓘ Azure for Students

Resource group * ⓘ ACR_AKS
[Create new](#)

Cluster details

Cluster preset configuration Standard (\$\$)
To quickly customize your Kubernetes cluster, choose one of the preset configurations above. You can modify these configurations at any time.
[Learn more and compare presets](#)

Kubernetes cluster name * ⓘ CISAКСDemoClr ✓

Region * ⓘ (US) Central US

Availability zones ⓘ Zones 1,2

[High availability is recommended for standard configurations](#)

[Review + create](#) [< Previous](#) [Next : Node pools >](#)

The Resource identity should be System-assigned managed identity.

Home > Create a resource > Kubernetes Service >

Create Kubernetes cluster ...

Basics Node pools **Access** Networking Integrations Advanced Tags Review + create

Resource identity ⓘ System-assigned managed identity
By default, Azure uses a managed identity. To use a service principal, use the CLI.
[Learn more](#)

Kubernetes authentication and authorization
Authentication and authorization are used by the Kubernetes cluster to control user access to the cluster as well as what the user may do once authenticated. [Learn more about Kubernetes authentication](#)

Role-based access control (RBAC) ⓘ ☒ Enabled ☐ Disabled

AKS-managed Azure Active Directory ⓘ ☐

[Review + create](#) [< Previous](#) [Next : Networking >](#)

Make sure that you check yes to Enable HTTP application routing.

Home > Create a resource > Kubernetes Service >

Create Kubernetes cluster

Network configuration ⓘ ☒ Kubenet ☐ Azure CNI

DNS name prefix * ⓘ CISAKSDemoClr-dns ✓

Traffic routing

Load balancer ⓘ Standard

Enable HTTP application routing ⓘ ☒

Security

Enable private cluster ⓘ ☐

Set authorized IP ranges ⓘ ☐

Network policy ⓘ ☒ None ☐ Calico ☐ Azure

[Review + create](#) [< Previous](#) [Next : Integrations >](#)

Select the previously created Container Registry and leave Container monitoring enabled.

Create Kubernetes cluster

Basics Node pools Access Networking Integrations Advanced Tags Review + create

Connect your AKS cluster with additional services.

Azure Container Registry
Connect your cluster to an Azure Container Registry to enable seamless deployments from a private image registry. You can create a new registry or choose one you already have. [Learn more about Azure Container Registry](#)

Container registry CIS325RegistryDemo ✓
[Create new](#)

Azure Monitor
In addition to the CPU and memory metrics included in AKS by default, you can enable Container Insights for more comprehensive data on the overall performance and health of your cluster. Billing is based on data ingestion and retention settings.
[Learn more about container performance and health monitoring](#)
[Learn more about pricing](#)

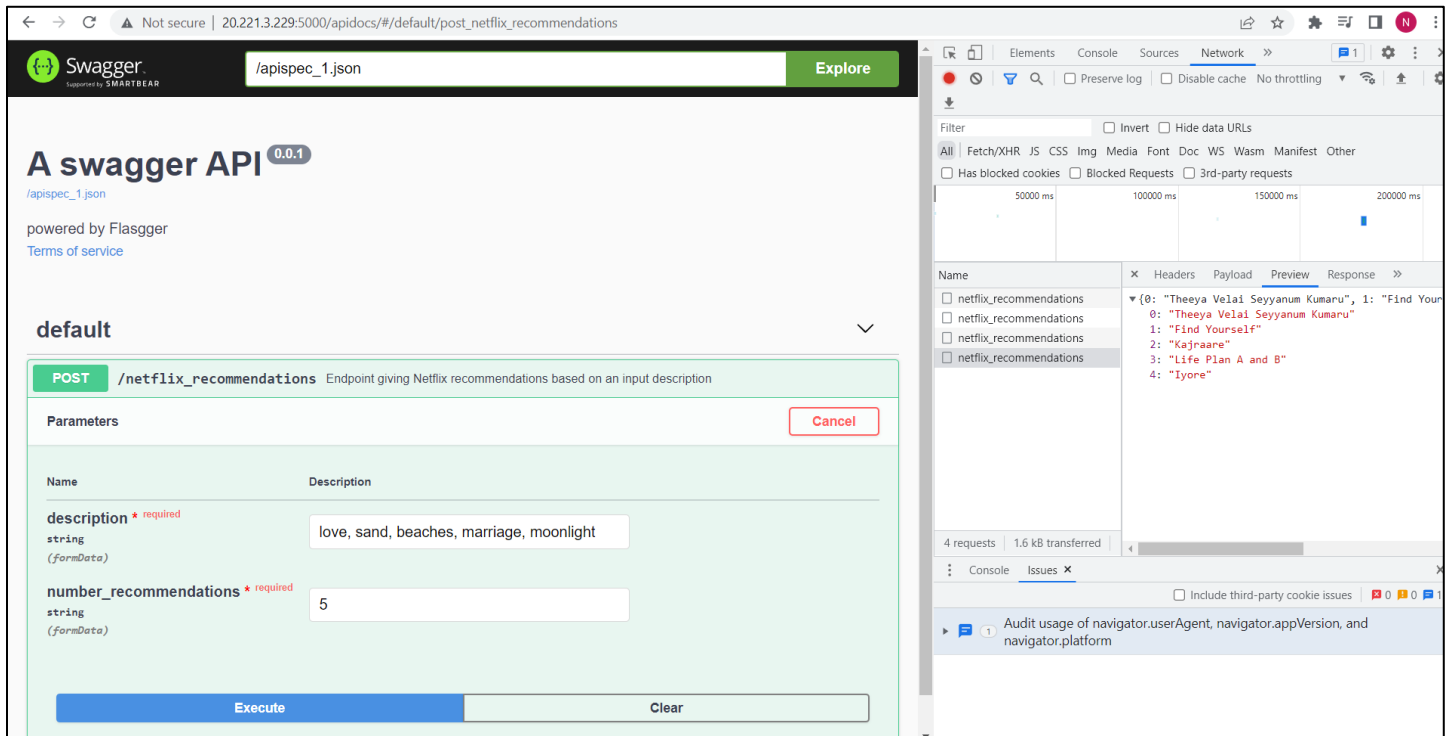
Container monitoring ☒ Enabled ☐ Disabled
 Azure monitor is recommended for standard configuration.

[Review + create](#) [< Previous](#) [Next : Advanced >](#)

Now you can review and create the cluster. Here is what the deployment confirmation will look like on command line interface:

```
C:\Users\nicho\CIS325\Final Project>kubectl get service netflix-recommender --watch
NAME                TYPE        CLUSTER-IP    EXTERNAL-IP    PORT(S)          AGE
netflix-recommender LoadBalancer 10.0.189.66    20.221.3.229   5000:32277/TCP   3m5s
```

Now the model can be publicly deployed as a web service. Using the “External IP”, you can log on to the user interface on any web browser. In this example, our public model testing IP is **20.221.3.220**.



The screenshot displays the Swagger API interface for a service running on 20.221.3.229:5000. The API is titled "A swagger API" and is powered by Flasgger. The selected endpoint is a POST request to /netflix_recommendations, which provides Netflix recommendations based on an input description. The form contains two required parameters: 'description' (a string) and 'number_recommendations' (a string). The 'description' field is filled with "love, sand, beaches, marriage, moonlight" and 'number_recommendations' is set to "5". The 'Execute' button is visible at the bottom of the form. On the right side, the browser's developer tools are open, showing the network tab with a single request. The response is a JSON array of 5 movie titles: ["Theeya Velai Seyyanum Kumaru", "Find Yourself", "Kajraare", "Life Plan A and B", "Iyore"].

As you can see above, a description was provided into the API with phrases and 5 titles were ‘recommended’ by the engine.

ALTERNATE RECOMMENDATION TECHNIQUES

An alternate method of creating a web-based recommendation engine was explored using cosine similarity to score and sort the spare matrix vectors for each title and then a recommendation function used to retrieve the 10 most similar titles. In this case the user can input just the title of the movie or show they liked and a list of similar titles would be returned. This proved to be a viable alternative method, with the majority of provided titles having similar genres, character and plot points. However, as this was not a true machine learning model with unique parameters that can be saved such as KNN or KMeans, it was not used for the API.

```
get_similar_titles('Midnight Mass')

7257    La Rosa de Guadalupe
1906                                Evil
3074                                Save Me
2971                                Diablero
4268                                Bird Box
7025                                Hungerford
2026                                Poshter Girl
8247    The Christmas Candle
2323                                The Mirror Boy
3352                                We Are the Wave
Name: title, dtype: object
```


ERROR & RESOLUTION LOG

It was discovered that there are problems with student accounts on Microsoft Azure. At the critical step selecting the node size to create the Kubernetes cluster, the system will not allow you to choose *any* size no matter the region, as shown below.

Home > Create a resource > Kubernetes Service >

Create Kubernetes cluster

Kubernetes version * ⓘ 1.22.6 (default) ▼

API server availability ⓘ

☒ 99.95%
Optimize for availability.

☐ 99.5%
Optimize for cost.

99.95% API server availability is recommended for standard configuration.

Primary node pool

The number and size of nodes in the primary node pool in your cluster. For production workloads, at least 3 nodes are recommended for resiliency. For development or test workloads, only one node is required. If you would like to add additional node pools or to see additional configuration options for this node pool, go to the 'Node pools' tab above. You will be able to add additional node pools after creating your cluster. [Learn more about node pools in Azure Kubernetes Service](#)

Node size * ⓘ

Standard DS2 v2
2 vcpus, 7 GiB memory

Standard DS2_v2 is recommended for standard configuration.

[Change size](#)

✖ This size is currently unavailable in this location for this subscription

Create Kubernetes cluster

✖ Validation failed. Click here to view details. →

Basics Node pools Access Networking Integrations Advanced Tags Review + create

Basics

Subscription	Azure for Students
Resource group	ACR_AKS
Region	Central US
Kubernetes cluster name	CISAKSDemoClr
Kubernetes version	1.22.6

Node pools

Node pools	1
Enable virtual nodes	Disabled

Summary Raw Error

ERROR DETAILS

Provisioning of resource(s) for container service CISAKSDemoClr in resource group ACR_AKS failed. Message: Category: ClientError; Code: Unspecified; SubCode: ; Message: Provisioning of resource(s) for container service CISAKSDemoClr in resource group ACR_AKS failed. Message: The template deployment failed with error: 'The resource with id: '/subscriptions/00750b85-dc9a-45c2-a798-3fcc2d354f33/resourceGroups/ACR_AKS/providers/Microsoft.Compute/virtualMachineScaleSets/aks-agentpool-32708500-vmss' failed validation with message: 'The requested size for resource '/subscriptions/00750b85-dc9a-45c2-a798-3fcc2d354f33/resourceGroups/ACR_AKS/providers/Microsoft.Compute/virtualMachineScaleSets/aks-agentpool-32708500-vmss' is currently not available in location 'centralus' zones '1,2' for subscription '00750b85-dc9a-45c2-a798-3fcc2d354f33'. Please try another size or deploy to a different location or zones. See https://aka.ms/azuresknotavailable for details.'. Details: ; InnerMessage: ; Dependency: ; OriginalError: %!s(<nil>); AKSTeam: . Details: (Code: Unspecified)

If you experience this problem, the only solution is to upgrade your account to the standard Pay As You Go account which will have the resources to obtain the proper resources. You may even have to create a brand new upgraded account to get it working.

GITHUB REPOSITORY

- ❖ <https://github.com/CIS325-Group2/MovieRecommenderSystem>

AUTHORS

- ❖ Nicholas Harrison
- ❖ Somi Poorey
- ❖ Kristy Seelnacht-Colombo

VERSION HISTORY

- ❖ 1.0 – Initial Release