

# LingoGym: A Language Learning Web Application

Independent Study Project Presentation

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April 11, 2022

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# Introduction

- A web application tool for English as a second/third language learners
- Target audience: Tibet and Ethiopia
- Use image recognition technologies

# Background and Motivation

- English: important for social mobility, opportunities, career development
- Difficulty with English learning among some groups of Tibetans
  - Elderly Tibetan refugees in India
  - Elderly Tibetan immigrants in the USA
  - Tibetans in Tibet
- Only 0.22% of Ethiopians speak English as of 2014 [4].

# Problem Statement and Objective

- Group of Tibetans/Ethiopians
  - Knows native spoken language only
  - Don't know how to read/write in their native language
  - Struggle with finding resources to learn language
  - Leads to difficulties in using Technology (mostly English-based)
- Objective
  - Create an inclusive web application
  - Flexible platform for users irrespective of their background
  - Image and keyword search facilitation

# Existing works

- English as a second language learning web sites
  - Manythings.org
  - BBC learning English website and app
  - Duolingo
  - Babbel
- What do they lack?

# What makes this project different?

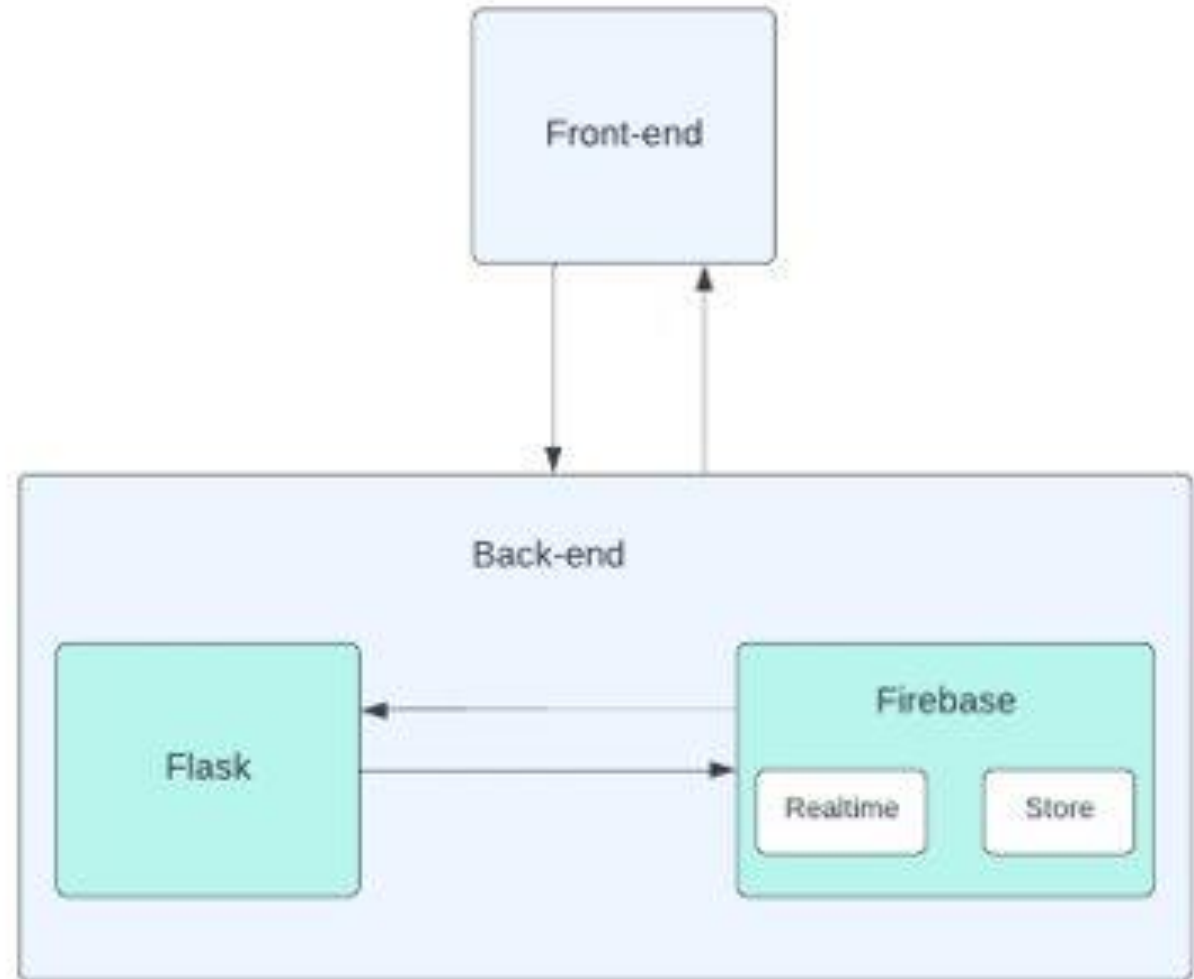
- Language learning tool
  - Targeted for Tibetan and Amharic
- Provide search by image feature
  - Content-based Image Retrieval technique
- Tibetan and Amharic focus
  - Both has less than 0.1% internet presence
  - Tibetan: No Google Translate support

# Project Components

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- Front-end
- Back-end
  - Flask
  - Firebase
    - Realtime database
    - Firebase Storage
- CMS

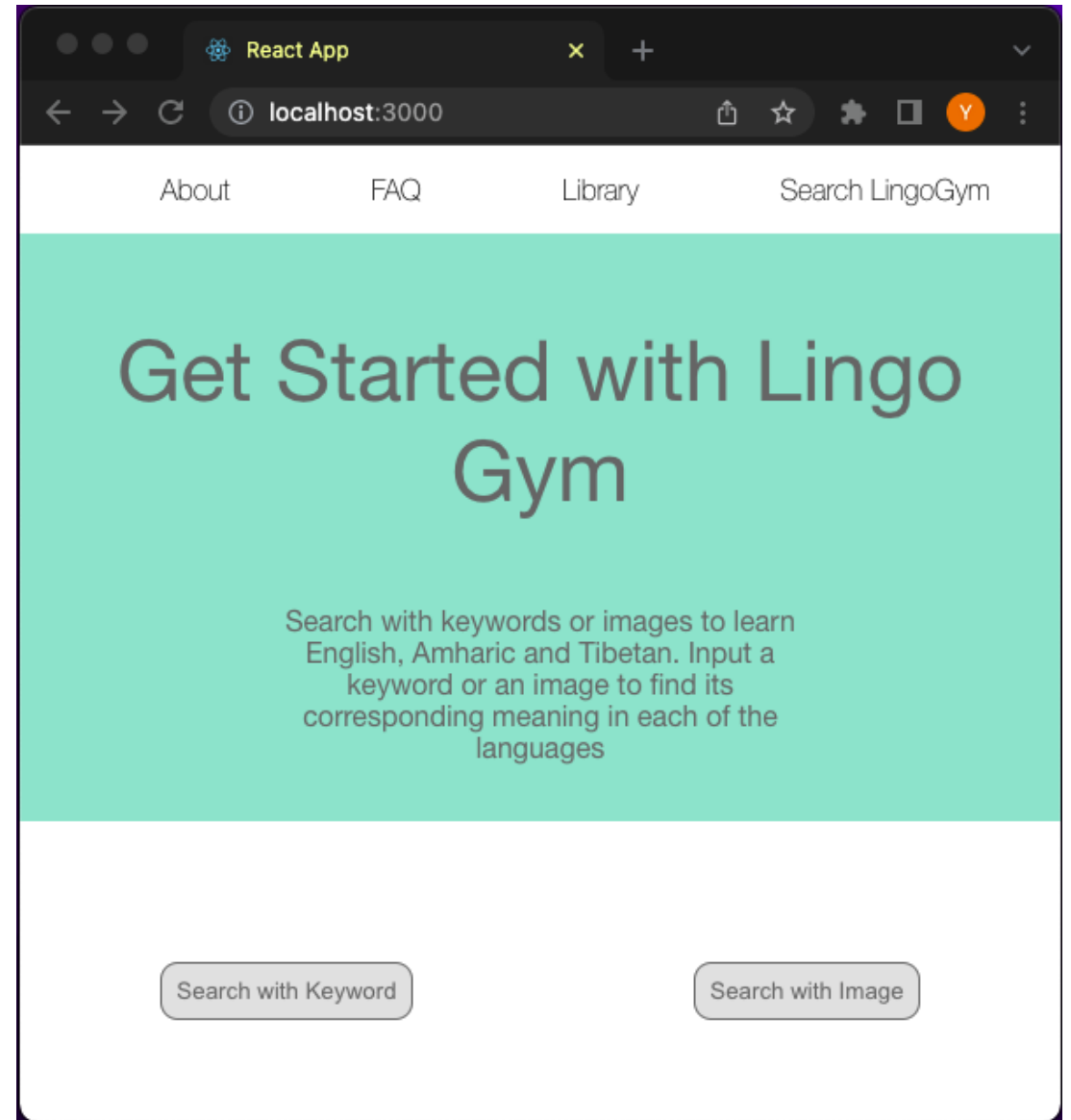
## Project Overview



# Front end

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- ReactJS
  - A JS library built and maintained by FB
- Flexible open-source library
- Industry giants: Apple, Netflix etc.

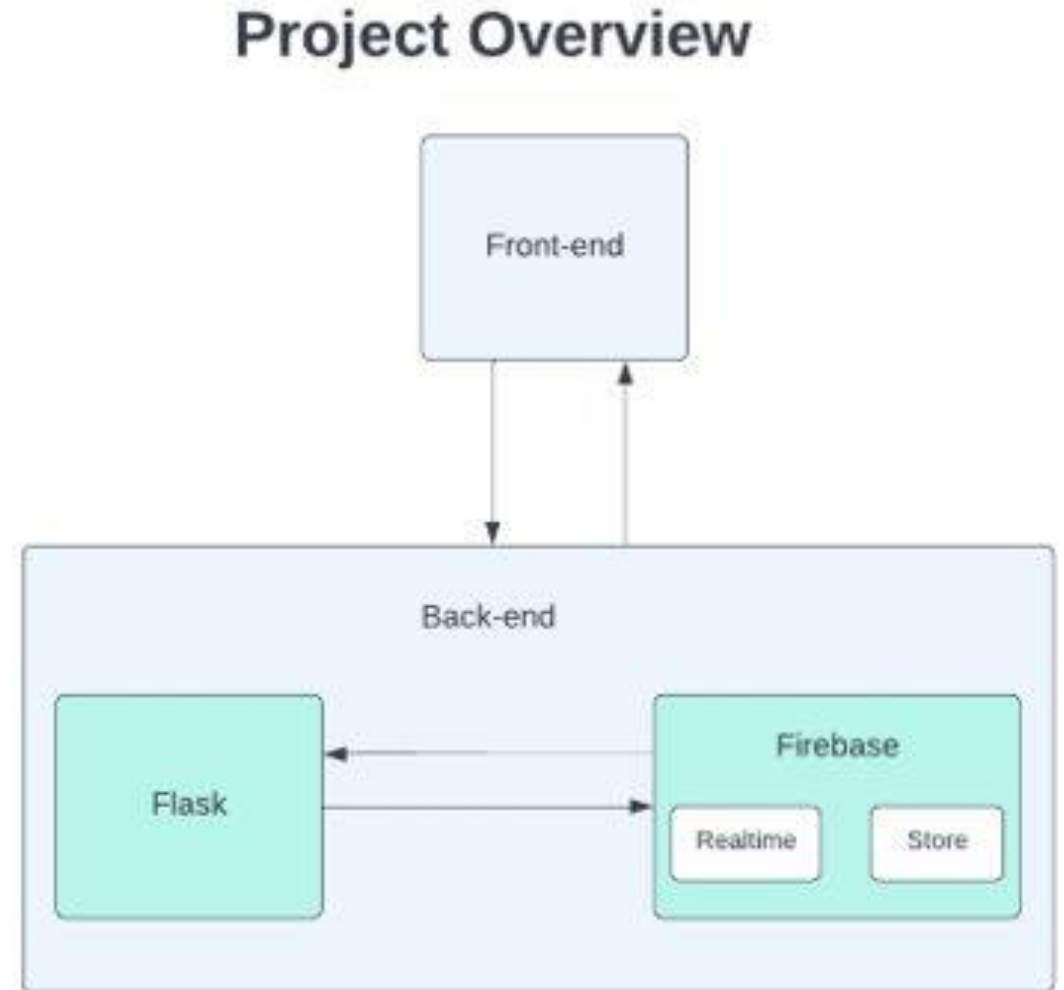




# Back end

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- Flask
  - Used for developing web app using python
- Firebase
  - Realtime database
  - Firebase storage



# Back end: Design decision factors

Flask:

- CBIR technique implementation
  - VGG16
  - Keras library from python

Firebase

- Realtime database
- Firebase storage
- Great API support
- Appropriate with the short development time

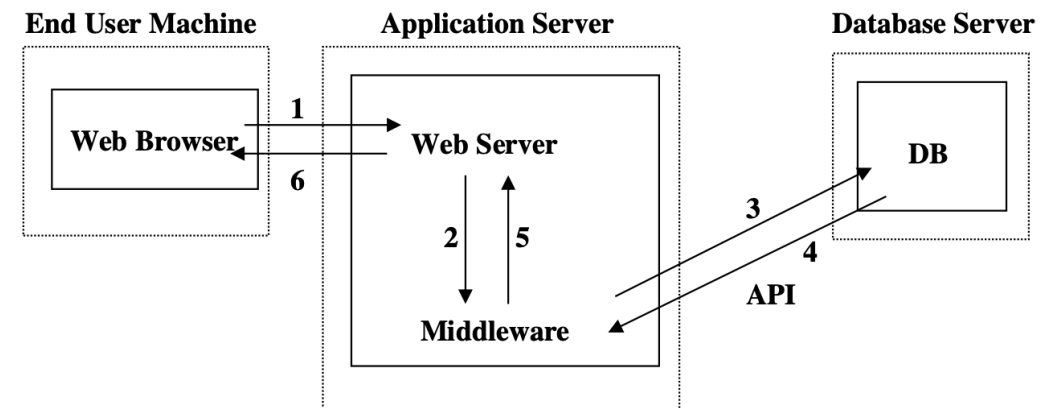


Figure 1. A generic web architecture

# Back-end: Realtime database

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1. Feature vectors
  - For each image in the data set
  - Key-value pair
  - Index, vector element
  - Used to process search by an image
2. Image tags and Info
  - Tag in Tibetan & Amharic
  - Link to the image in fire storage
  - Usable for API calls to perform CRUD on the data set.

cs5990sp22-default-rtdb

## Features

```
cat: "{0: 0.0, 1: 0.0, 2: 0.0, 3: 0.0, 4: 0.0, 5: 0.0...}"  
dog: "{0: 0.106378004, 1: 0.0, 2: 0.0, 3: 0.0, 4: 0.0...}"
```

## Imgs

### cat

```
amharic: "ፍጥረት"  
english: "cat"  
link: "https://firebasestorage.googleapis.com/v0/b/cs5..."  
tibetan: "ཁྲིམ་པ་"
```

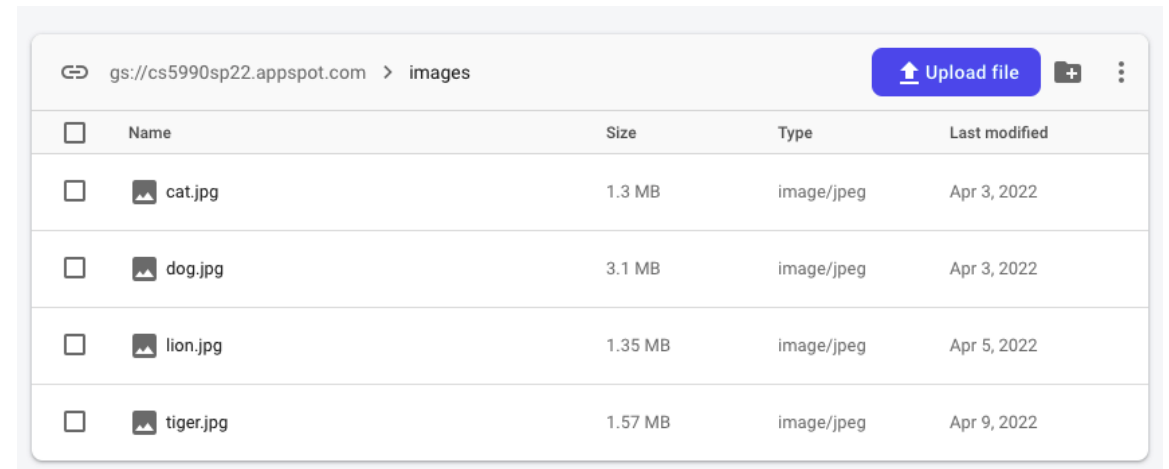
### dog

```
amharic: "ወሽ"   
english: "dog"  
link: "https://firebasestorage.googleapis.com/v0/b/cs5..."  
tibetan: "ཐྱི་"
```





# Back – end: Firebase Storage

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- Stores media: files, images
- Links generated for each image
- Accessible via api-calls from Realtime database



The screenshot shows the Firebase Storage console interface. At the top, the breadcrumb path is 'gs://cs5990sp22.appspot.com > images'. There is an 'Upload file' button and a menu icon. Below this is a table listing the stored files.

<input type="checkbox"/>	Name	Size	Type	Last modified
<input type="checkbox"/>	 cat.jpg	1.3 MB	image/jpeg	Apr 3, 2022
<input type="checkbox"/>	 dog.jpg	3.1 MB	image/jpeg	Apr 3, 2022
<input type="checkbox"/>	 lion.jpg	1.35 MB	image/jpeg	Apr 5, 2022
<input type="checkbox"/>	 tiger.jpg	1.57 MB	image/jpeg	Apr 9, 2022

# Major features of the project

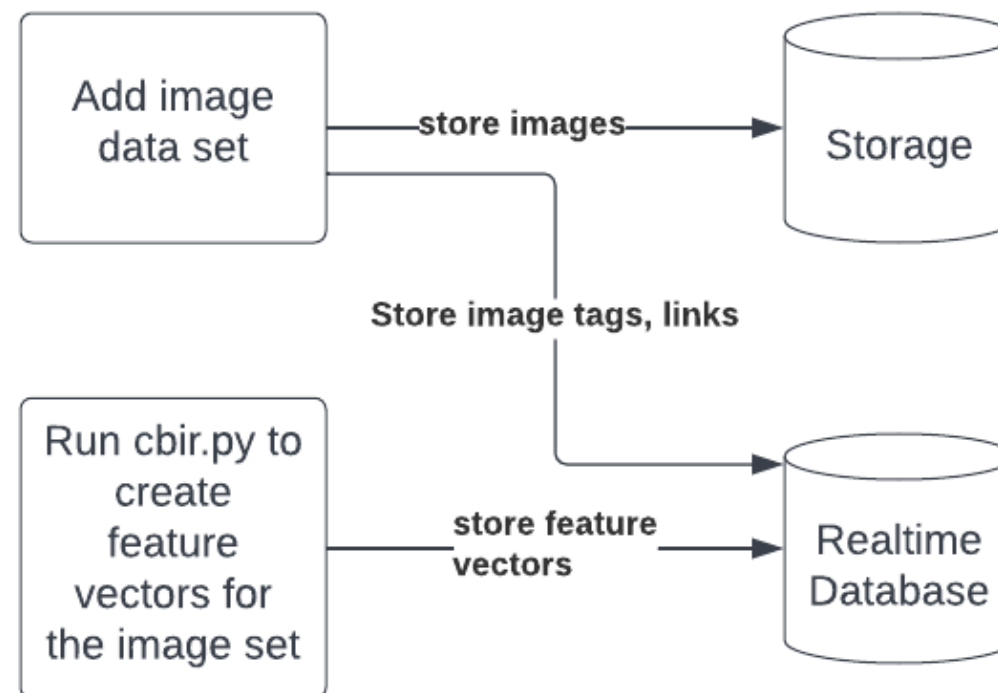
- Keyword based Content Retrieval
- Image based Content Retrieval
- Content Management Site

# Database Preparation

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- Content Management Site
  - A separate program
  - Run on its own
- Upload data sets
  1. captions for image-tagging
  2. Image data set

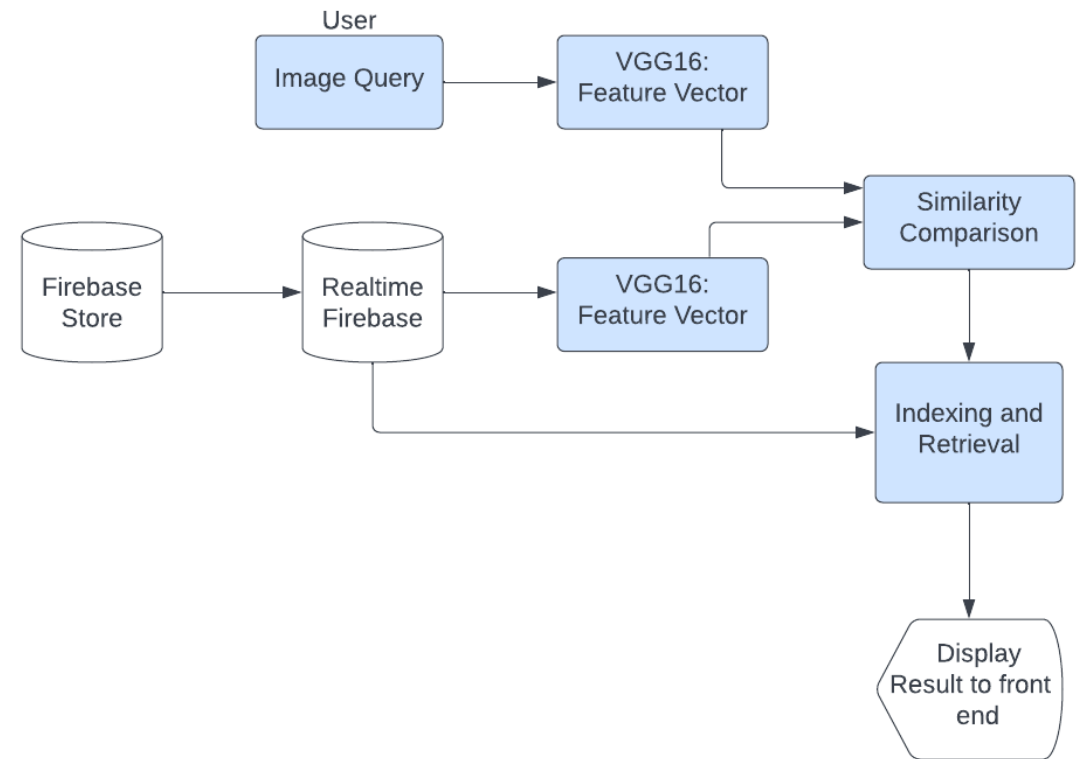
## Content Management Site



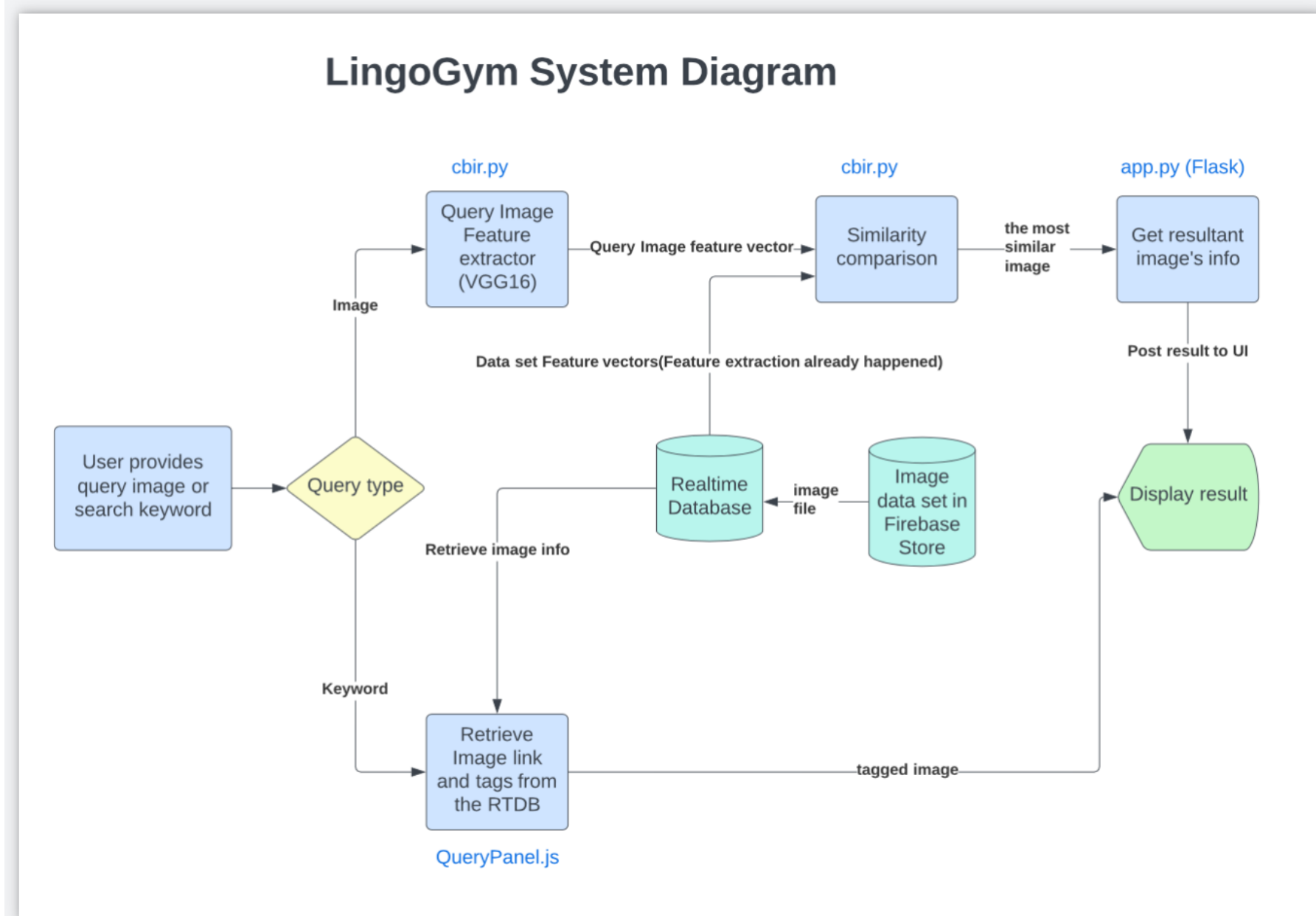
# Content-based Image Retrieval

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- A technique to query an image over a database of images to get a set of results that most closely matches the queries image.
- VGG 16
  - A pre-trained deep CNN model
- Ease of implementation and model accuracy



# Connecting project components

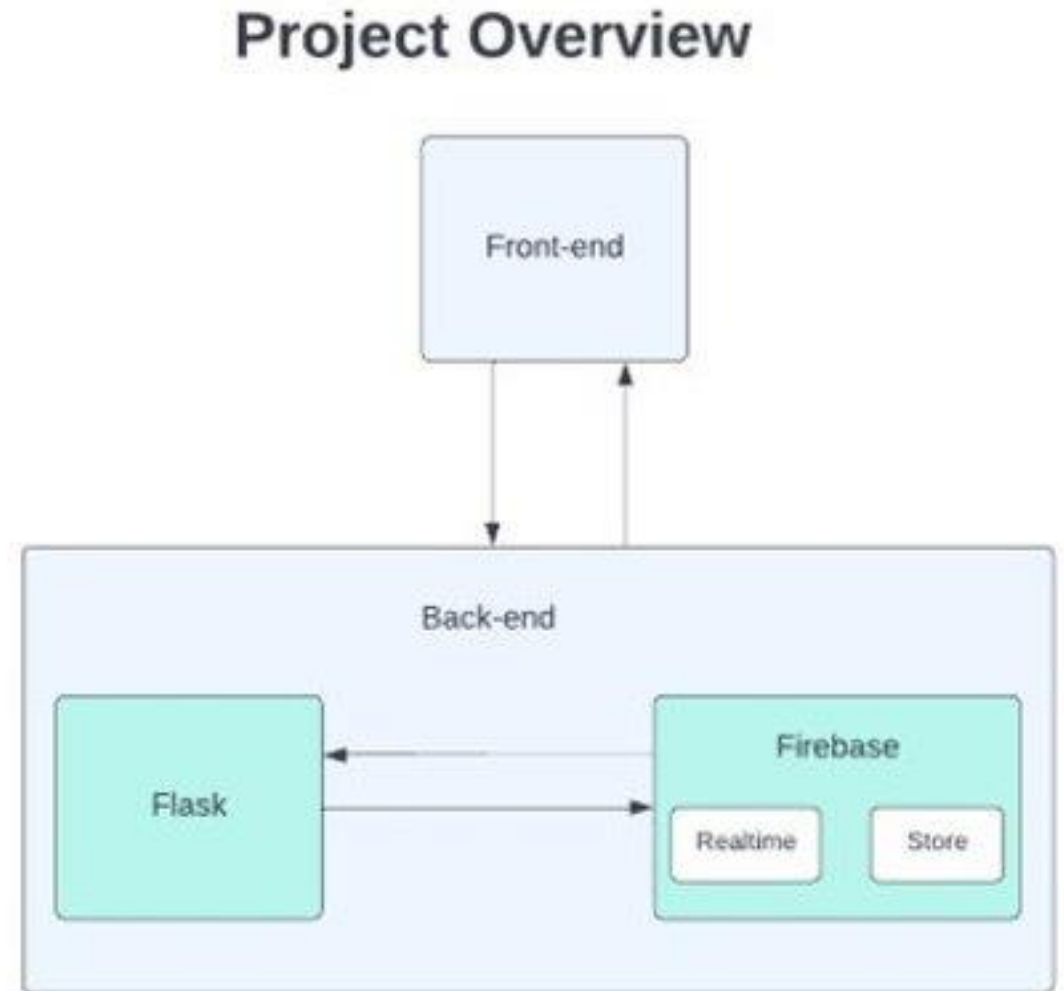




# Main Contributions

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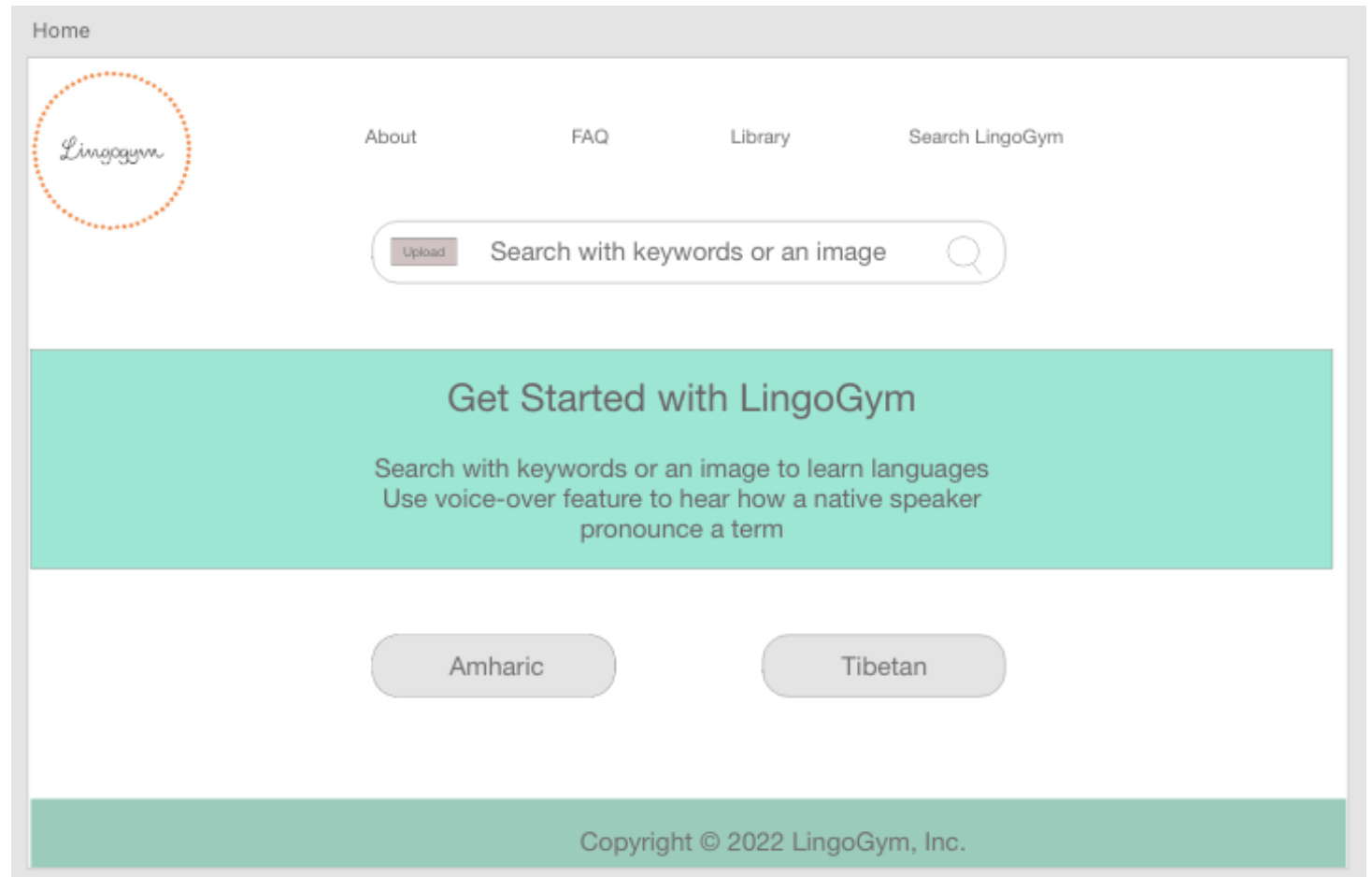
- Communication between flask and firebase
- Implement CBIR using VGG 16
- Find an efficient way to retrieve similar images
  - Store feature vectors for each image in the data set to realtime database
  - Improve retrieval speed
  - Extract feature once but api-calls needed to access the feature vectors.



# Contribution: Front end UI Design

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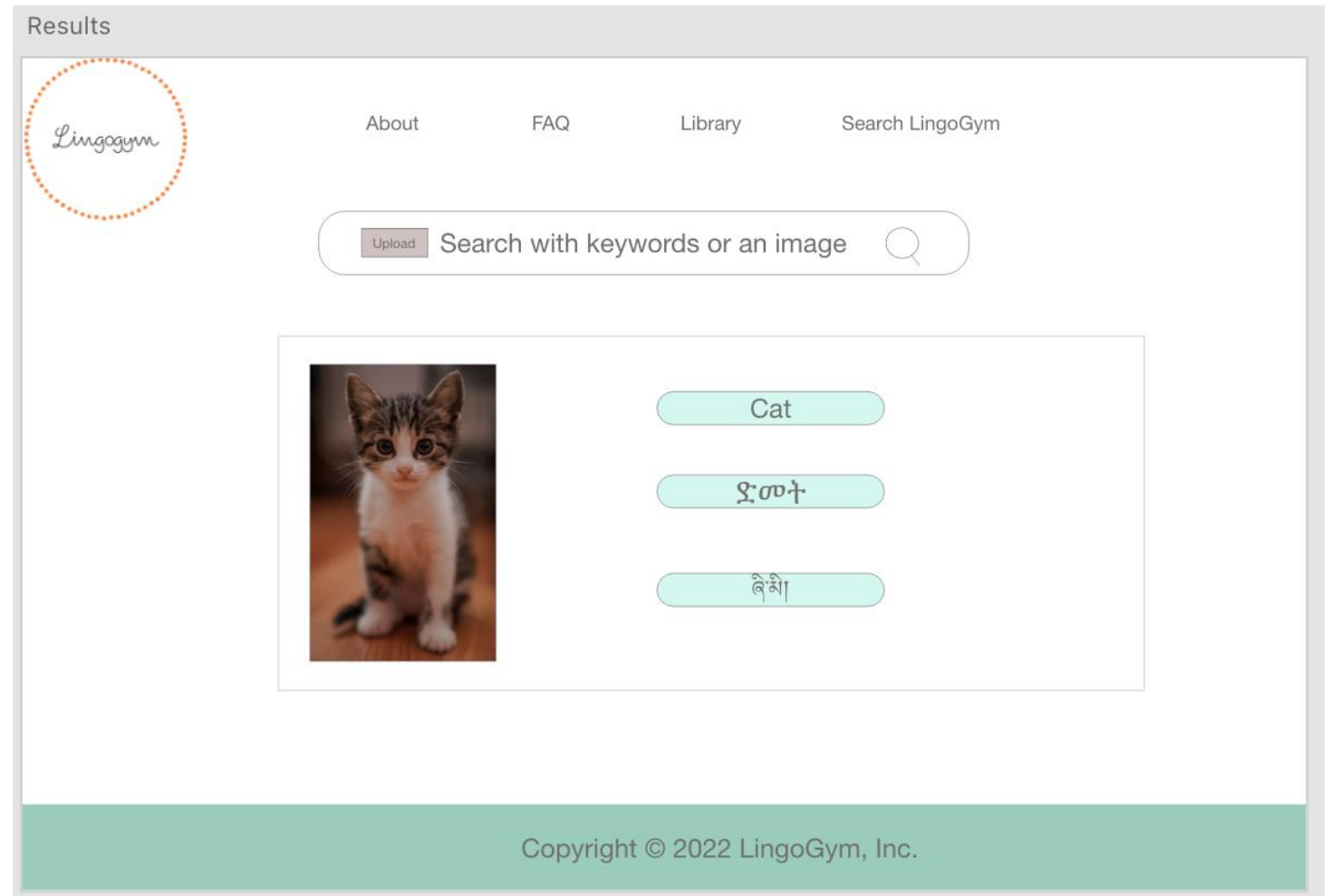
- Adobe XD
- Support from OU IT Software
- Generate links to be shared with teammate
- Easy to use



# Contribution: Front end UI Design

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- UI displaying output



# Communication between Flask and Firebase

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- Pyrebase interface
- Communicate with Firebase's REST API.
- Installation
  - Pipenv install pyrebase

```
config = {  
    'apiKey': "AIzaSyBazt2gIqkdNpvU8e4t_oJD-BNq0t0towM",  
    'authDomain': "cs5990sp22.firebaseio.com",  
    'databaseURL': "https://cs5990sp22-default-rtdb.firebaseio.com",  
    'projectId': "cs5990sp22",  
    'storageBucket': "cs5990sp22.appspot.com",  
    'messagingSenderId': "622617212731",  
    'appId': "1:622617212731:web:a9cf206aa33159a292291e",  
    'measurementId': "G-0YHMPKFEQW"  
}
```

# Firestore CRUD Operations using python

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Create Feature Vectors for  
image data set and store in  
Realtime database

```
def put_feature_data():
    firebase = pyrebase.initialize_app(config)
    db = firebase.database()
    imgs = db.child("Imgs").get()
    fe = FeatureExtractor()
    all_features = {}
    for e in imgs.each():
        individual_feature = {}
        val = e.val()
        key = e.key()
        link = val["link"]
        feature = fe.extract(link)
        c = 0
        for (variable) individual_feature: dict
            individual_feature[c] = x
            c = c+1
        print(key)
        all_features[key] = str(individual_feature)
    db.child("Features").set(all_features)
```

# Firestore CRUD Operations using python

---

Read Feature Vectors for  
image data set stored  
in Realtime database

For similarity comparison

```
def get_feature_data():  
    firebase = pyrebase.initialize_app(config)  
    db = firebase.database()  
    feature_list = []  
    features = db.child("Features").get()  
    for f in features.each():  
        val = f.val()  
        # feature values from dictionary values  
        val_dict = ast.literal_eval(val)  
  
        # change to float  
        values = [float(x) for x in list(val_dict.values())]  
        # ndarray of feature vectors  
        f_vec = np.array(values)  
  
        # add to feature list for similarity computation  
        feature_list.append(f_vec)  
    return feature_list
```

# Query Image's Feature Vector

Apply VGG 16 on the query image to extract its feature vector.

For similarity comparison with the feature vectors for images from dataset

```
def get_feature_QI (img_qi):  
    fe = FeatureExtractor()  
    return fe.extractQI(img_qi)
```

# Similarity Comparison

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Subtract query image's feature vector from that of the feature vectors for images in the dataset.

Return information about the most matching feature vector to that of the QI

Includes image tags and link

```
def get_similar_img(img_QI):
    firebase = pyrebase.initialize_app(config)
    db = firebase.database()
    imgs = db.child("Imgs").get()
    image_data = []
    for e in imgs.each():
        val = e.val()
        # key = e.key()
        # print(val)
        # link = val["link"]
        image_data.append(val)
    feature_set = get_feature_data()
    feature_QI = get_feature_QI(img_QI) # feature_set[0]
    dists = np.linalg.norm(feature_set-feature_QI, axis=1)
    rank_index = np.argsort(dists)[:1]
    scores = [(dists[id], image_data[id]) for id in rank_index]
    scores[0][1].update({"scores":scores[0][0]})
    #print("***** Scores ", scores[0][1])
    # the most similar image to the QI
    return scores[0][1]
```



# CBIR

## Implementation:

### VGG 16

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- Feature Extractor method for the image dataset

```
def extract(self, url):  
  
    response = requests.get(url)  
    img = Image.open(BytesIO(response.content))  
    # print (img)  
    img = img.resize((224, 224)) # VGG must take a 224x224 img as an input  
    img = img.convert('RGB') # Make sure img is color  
    x = image.img_to_array(img) # To np.array. Height x Width x Channel. dtype=float32  
    x = np.expand_dims(x, axis=0) # (H, W, C)->(1, H, W, C), where the first elem is the number of img  
    x = preprocess_input(x) # Subtracting avg values for each pixel  
    feature = self.model.predict(x)[0] # (1, 4096) -> (4096, )  
    return feature / np.linalg.norm(feature) # Normalize
```

# CBIR

## Implementation: VGG 16

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- Feature Extractor method for the query image

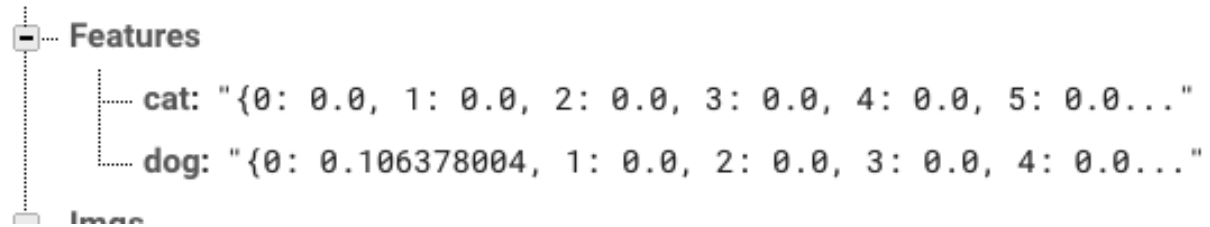
```
def extractQI(self, img):  
  
    # print (img)  
    img = img.resize((224, 224)) # VGG must take a 224x224 img as an input  
    img = img.convert('RGB') # Make sure img is color  
    x = image.img_to_array(img) # To np.array. Height x Width x Channel. dtype=float32  
    x = np.expand_dims(x, axis=0) # (H, W, C)->(1, H, W, C), where the first elem is the number of img  
    x = preprocess_input(x) # Subtracting avg values for each pixel  
    feature = self.model.predict(x)[0] # (1, 4096) -> (4096, )  
    return feature / np.linalg.norm(feature) # Normalize
```

# Issues faced

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- How to store feature vectors in Realtime database?
  - Firebase: no native support for arrays
  - Data stored as key-value pairs
- Solution
  - Key-value pairs created for each feature vector
  - Index as key, element as value
  - Feature vectors successfully stored in Firebase
- Requirements
  - Read feature vectors: make API call to firebase
  - Similarity comparison: convert feature vectors to NumPy array first

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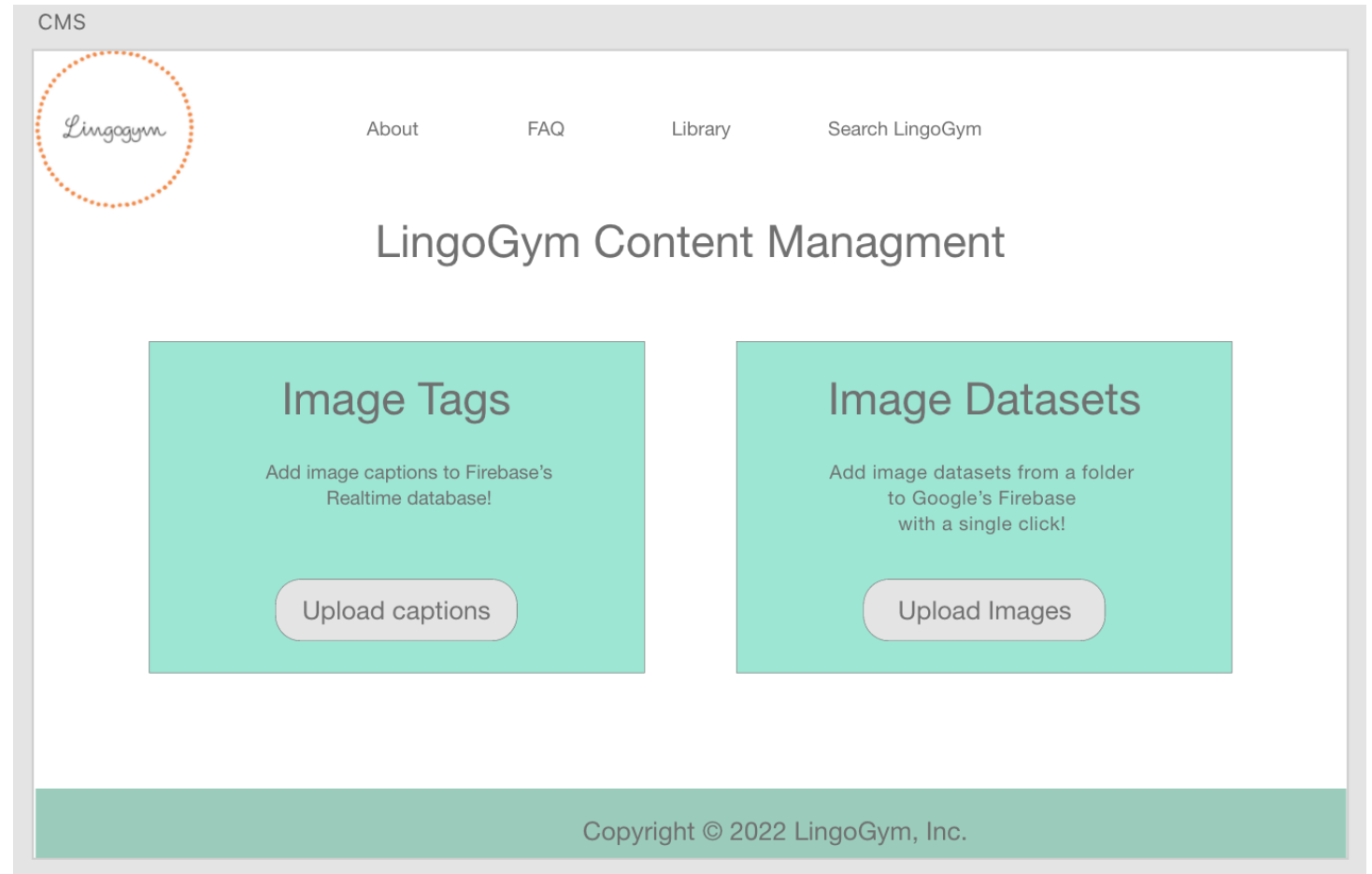


Demo

# Conclusion and Future Work

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- Per project objective
  - Have been able to create a web application
  - Retrieve similar images
  - Multilingual search
- Future Work
  - Voice over feature not integrated yet
  - Increase image dataset to cover more diverse and advanced topics
  - Explore other CBIR techniques
  - Integrate more languages
  - Implementation of CMS UI



# References

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