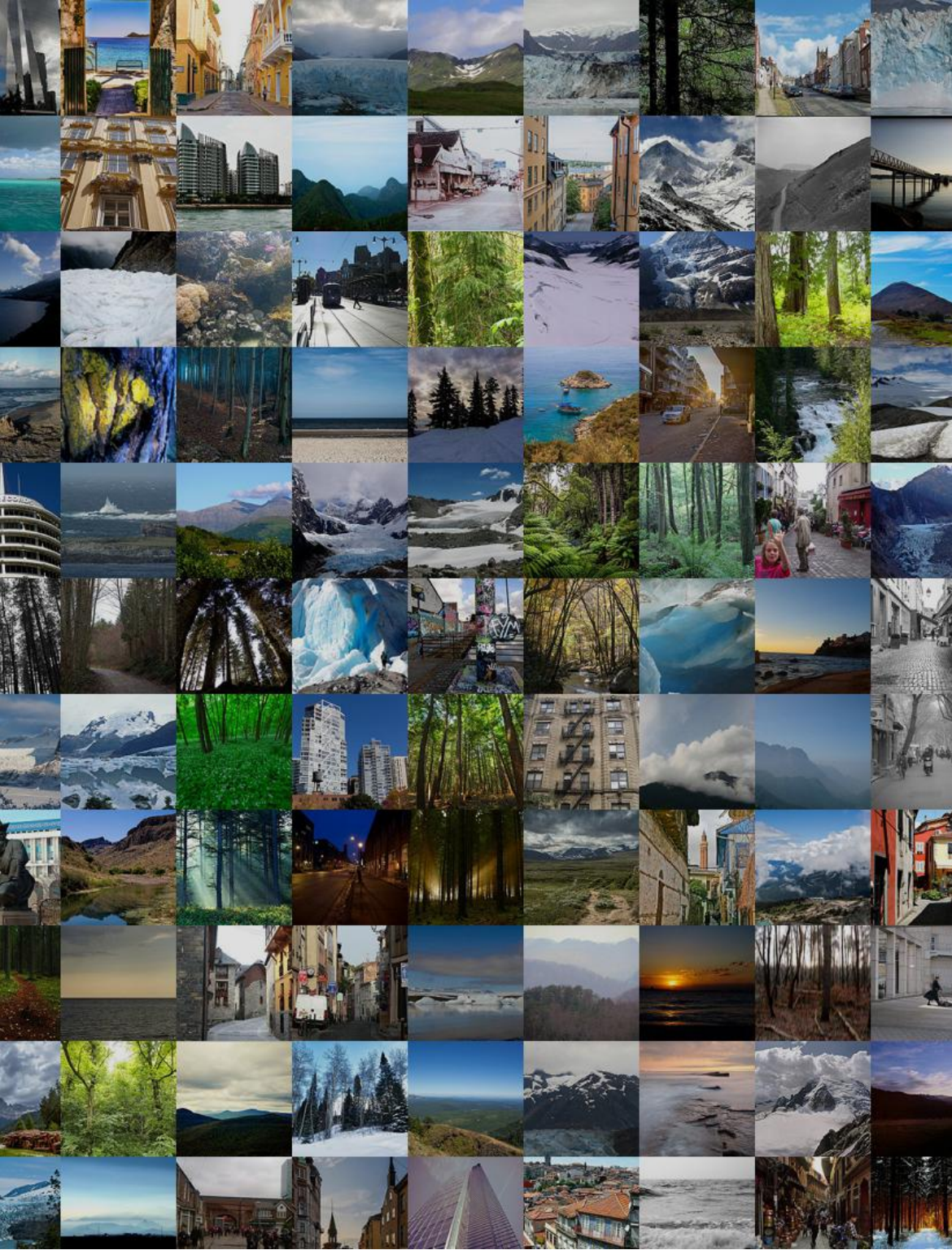




Image Archiving Model

Date:
06/27/2023

Presented by:
Ali Hijazy



Objective

Create a model capable of distinguishing between six distinct types of scenes.

Use

- Streamlines the image archiving workflow by automating the classification task.

Operational Benefits

- Significantly reduces the time required for image classification during the archiving process
- Enables better searchability and retrieval of archived images based on scene type.



street



street



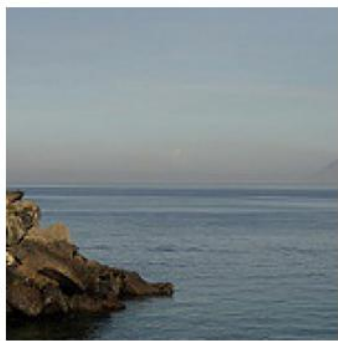
buildings



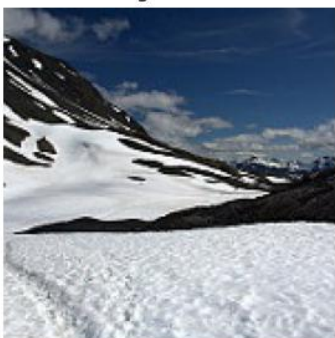
mountain



street



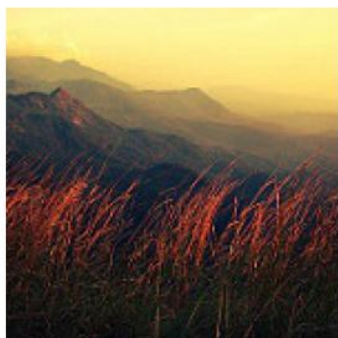
sea



glacier



street



mountain



glacier



forest



buildings

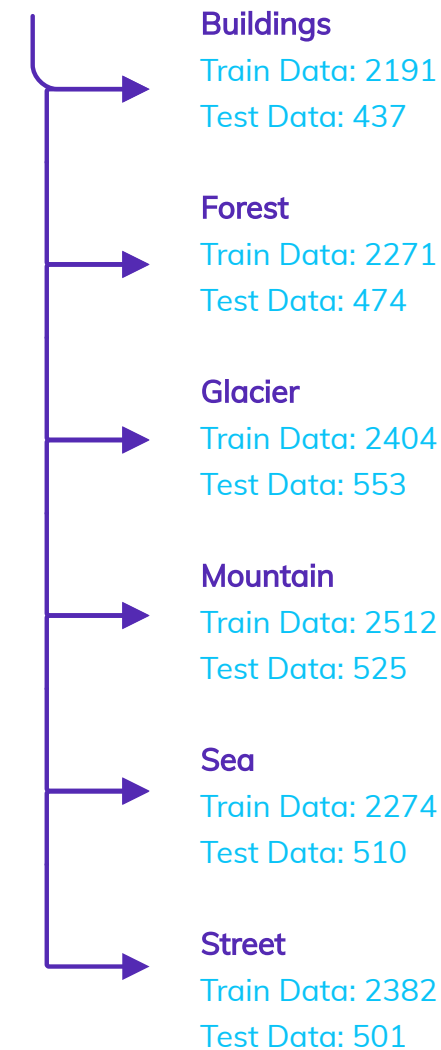
DATASET

kaggle

The dataset is
17000 labeled
images divided into
6 classes

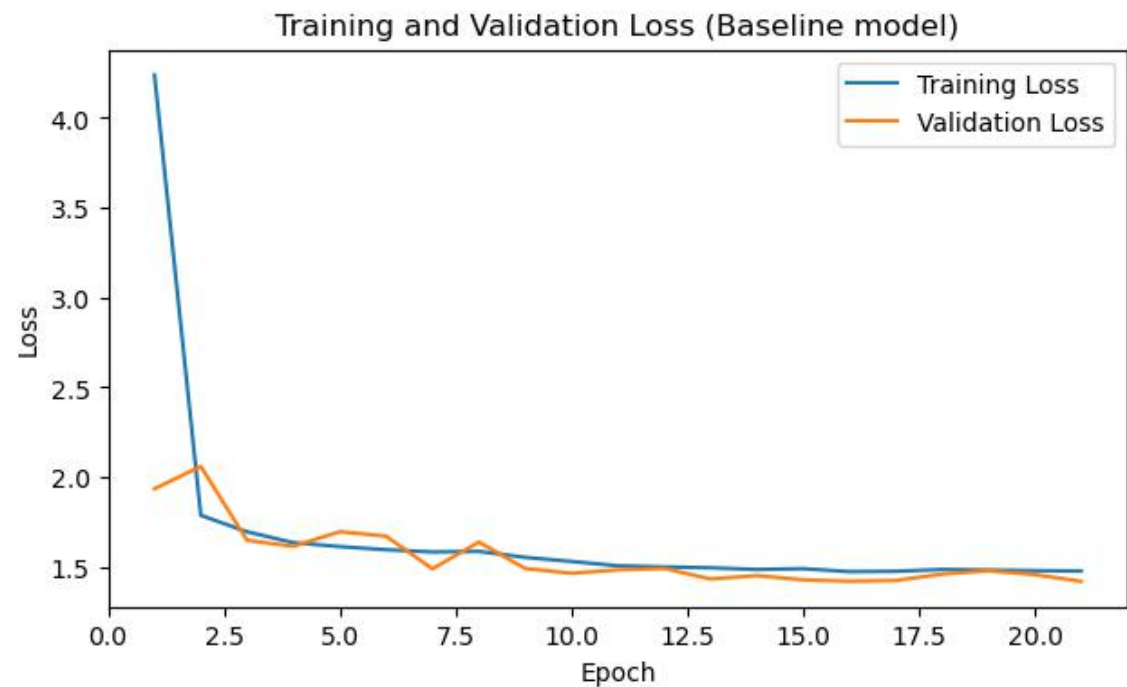
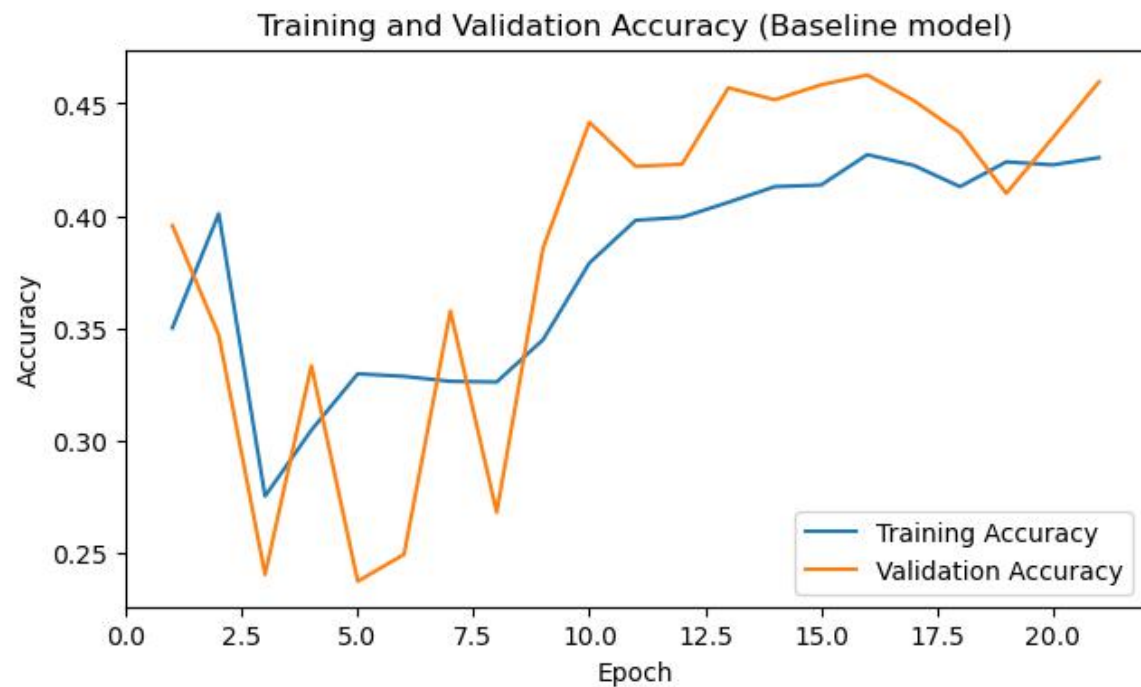
images are colored
 $150 * 150$ pixels

Target classes



Baseline Model (Neural network)

Test: 0.46
Train: 0.43

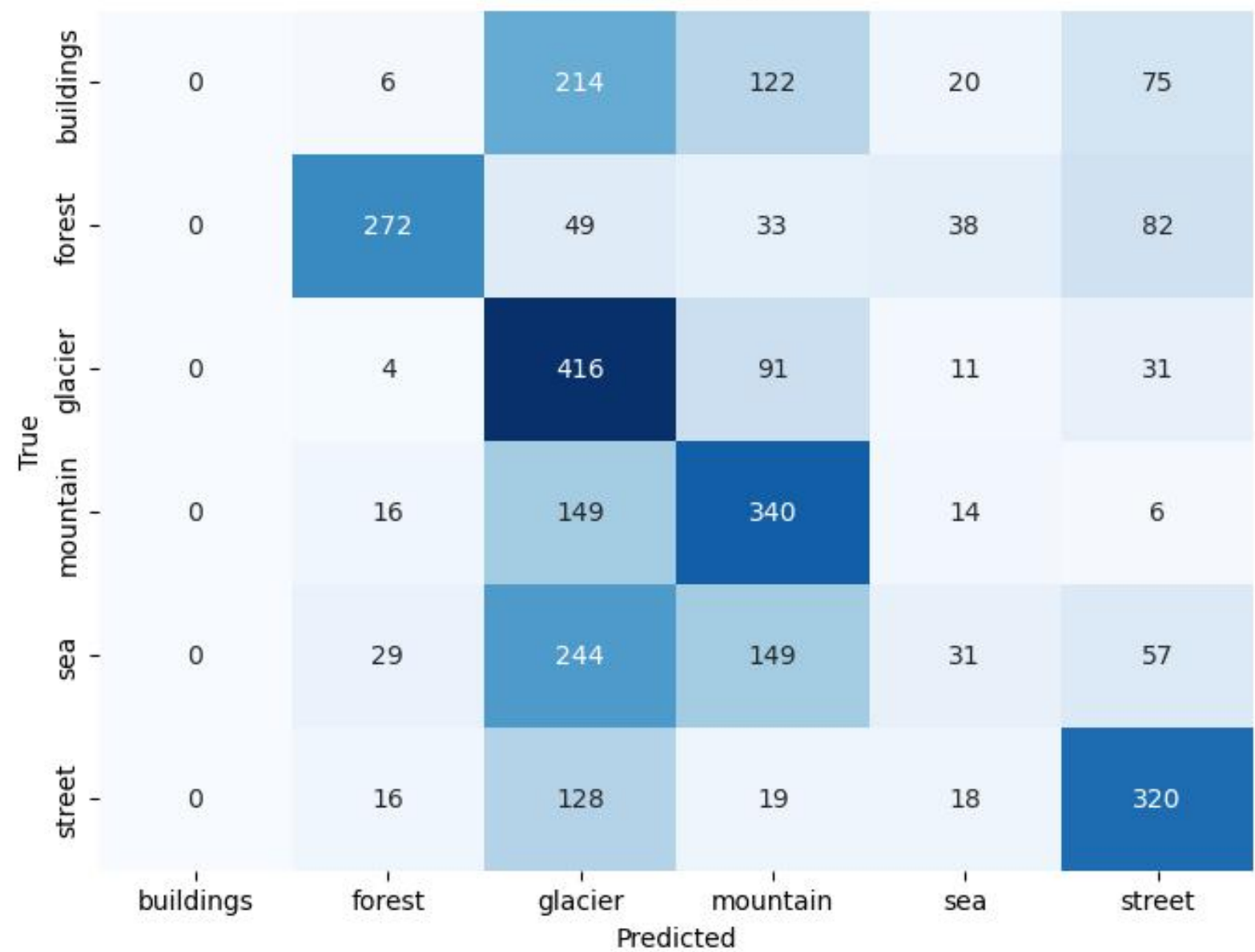


Input Layer (Flatten)
Hidden Layer (Dense with 128 neurons)
Output Layer (Dense with 6 neurons)

Baseline Model (Neural network)

Test: 0.46
Train: 0.43

Confusion Matrix



True: glacier Predicted: glacier



True: glacier Predicted: glacier



True: buildings Predicted: glacier



True: street Predicted: glacier



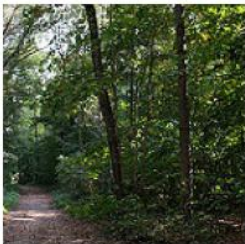
True: sea Predicted: glacier



True: street Predicted: glacier



True: forest Predicted: forest



True: buildings Predicted: glacier



True: street Predicted: street



True: street Predicted: glacier



True: forest Predicted: glacier



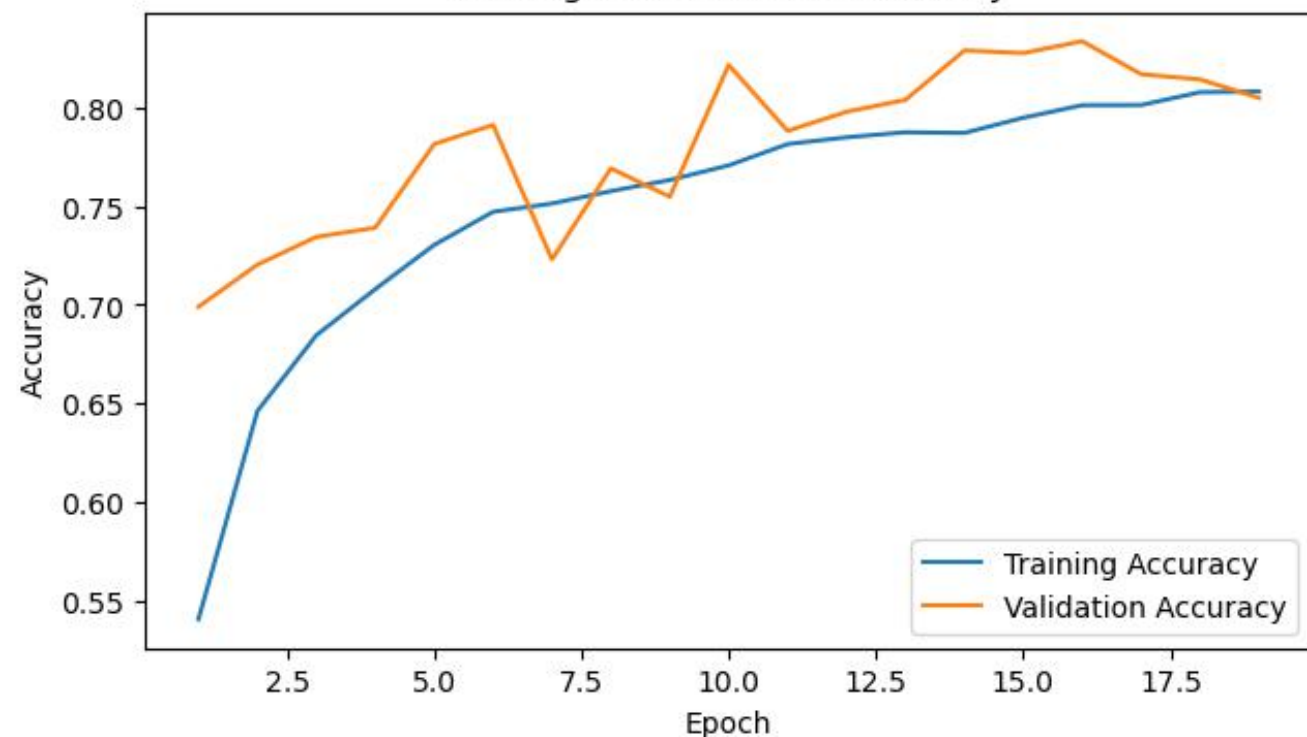
True: glacier Predicted: street



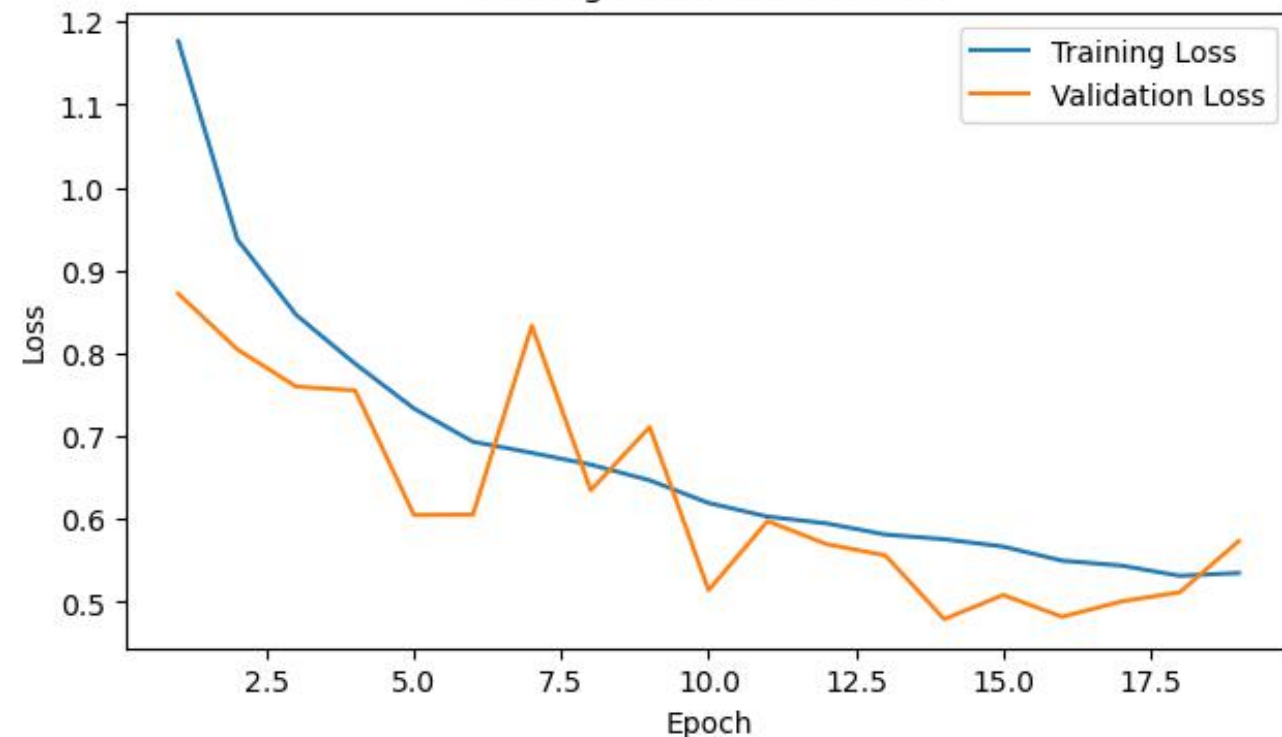
Model 2 (3 Convolutional Layers)

Test: 0.80
Train: 0.81

Training and Validation Accuracy



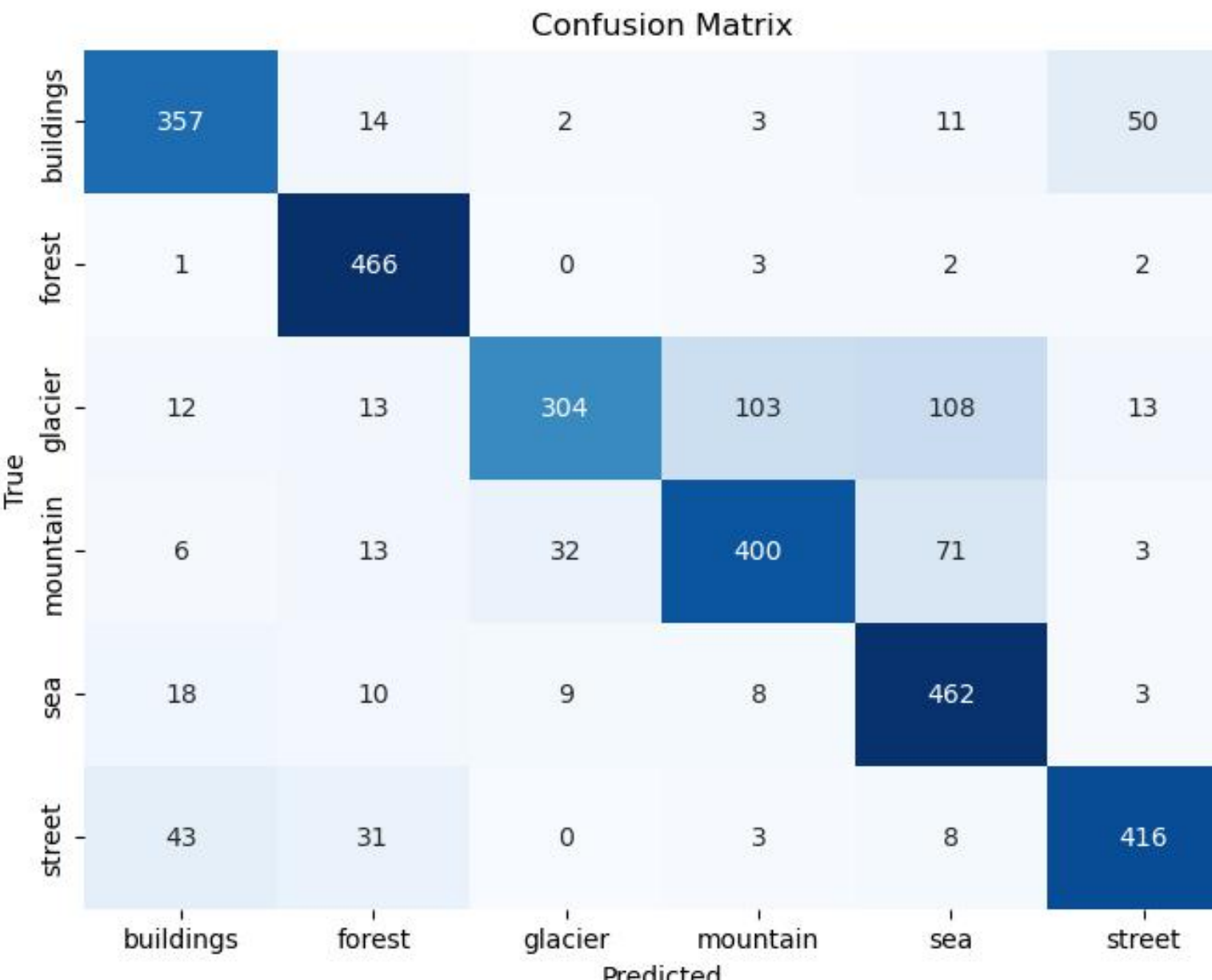
Training and Validation Loss



Conv2D(filters=32, kernel_size=(3, 3))
MaxPooling2D(pool_size=(3, 3))
Conv2D(filters=64, kernel_size=(3, 3))
MaxPooling2D(pool_size=(2, 2))
Conv2D(filters=128, kernel_size=(3, 3))
Flatten()
Dense(filters=128)
Dense(filters=6)

Model 2 (2 Convolutional Layers)

Test: 0.80
Train: 0.81



True: street Predicted: mountain



True: sea Predicted: sea



True: mountain Predicted: mountain



True: sea Predicted: sea



True: mountain Predicted: sea



True: buildings Predicted: buildings



True: forest Predicted: forest



True: buildings Predicted: street



True: sea Predicted: sea



True: glacier Predicted: glacier



True: sea Predicted: sea



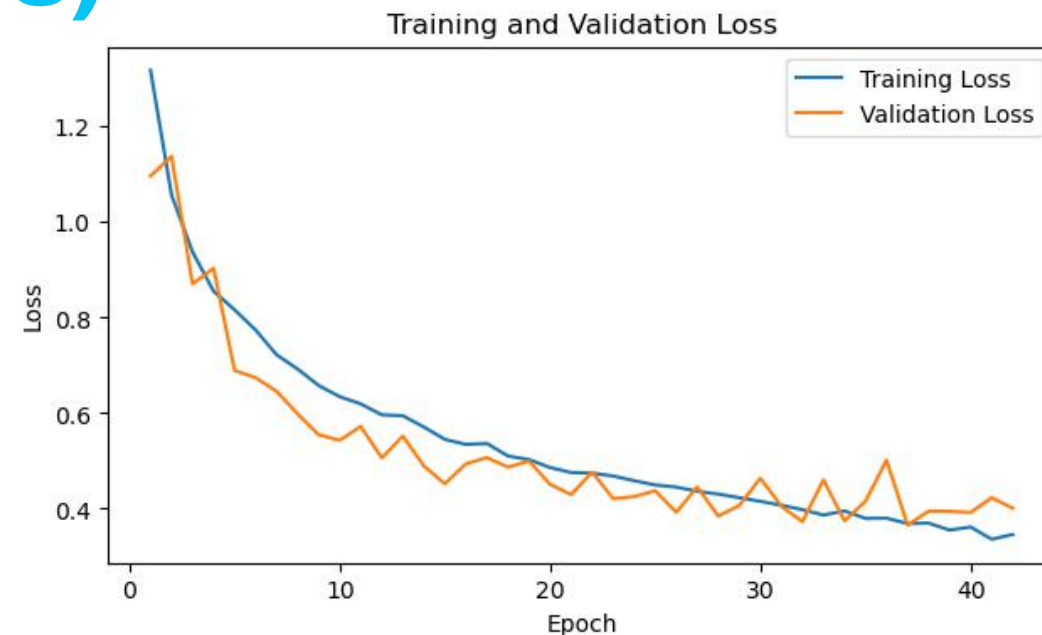
True: mountain Predicted: mountain



Model 3 (6 Convolutional Layers)

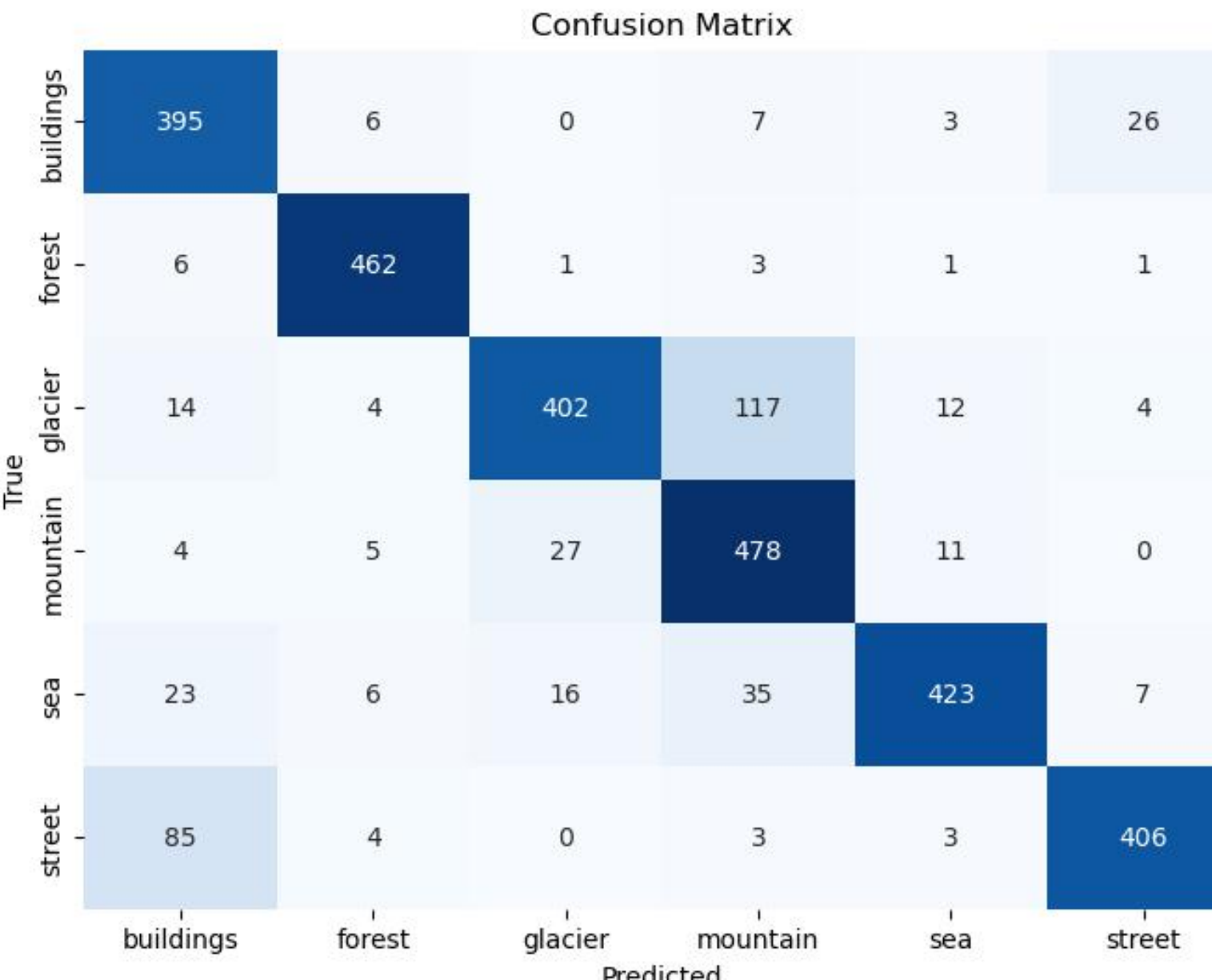
Test: 0.86
Train: 0.87

```
Conv2D(filters=32, kernel_size=(3, 3))  
MaxPooling2D(pool_size=(3, 3))  
Conv2D(filters=64, kernel_size=(3, 3))  
MaxPooling2D(pool_size=(2, 2))  
Conv2D(filters=128, kernel_size=(3, 3))  
MaxPooling2D(pool_size=(2, 2))  
Conv2D(filters=256, kernel_size=(3, 3))  
MaxPooling2D(pool_size=(2, 2))  
Conv2D(filters=512, kernel_size=(3, 3))  
Conv2D(filters=1024, kernel_size=(3, 3))  
Flatten()  
Dense(filters=1024)  
Dense(filters=6)
```



Model 3 (6 Convolutional Layers)

Test: 0.86
Train: 0.87



True: sea Predicted: sea



True: forest Predicted: sea



True: glacier Predicted: sea



True: forest Predicted: sea



True: mountain Predicted: sea



True: buildings Predicted: sea



True: sea Predicted: sea



True: sea Predicted: sea



True: buildings Predicted: sea



True: sea Predicted: sea



True: mountain Predicted: sea

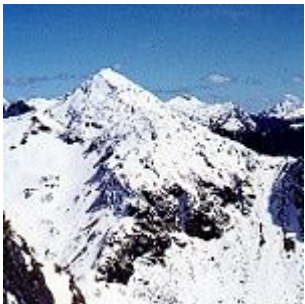
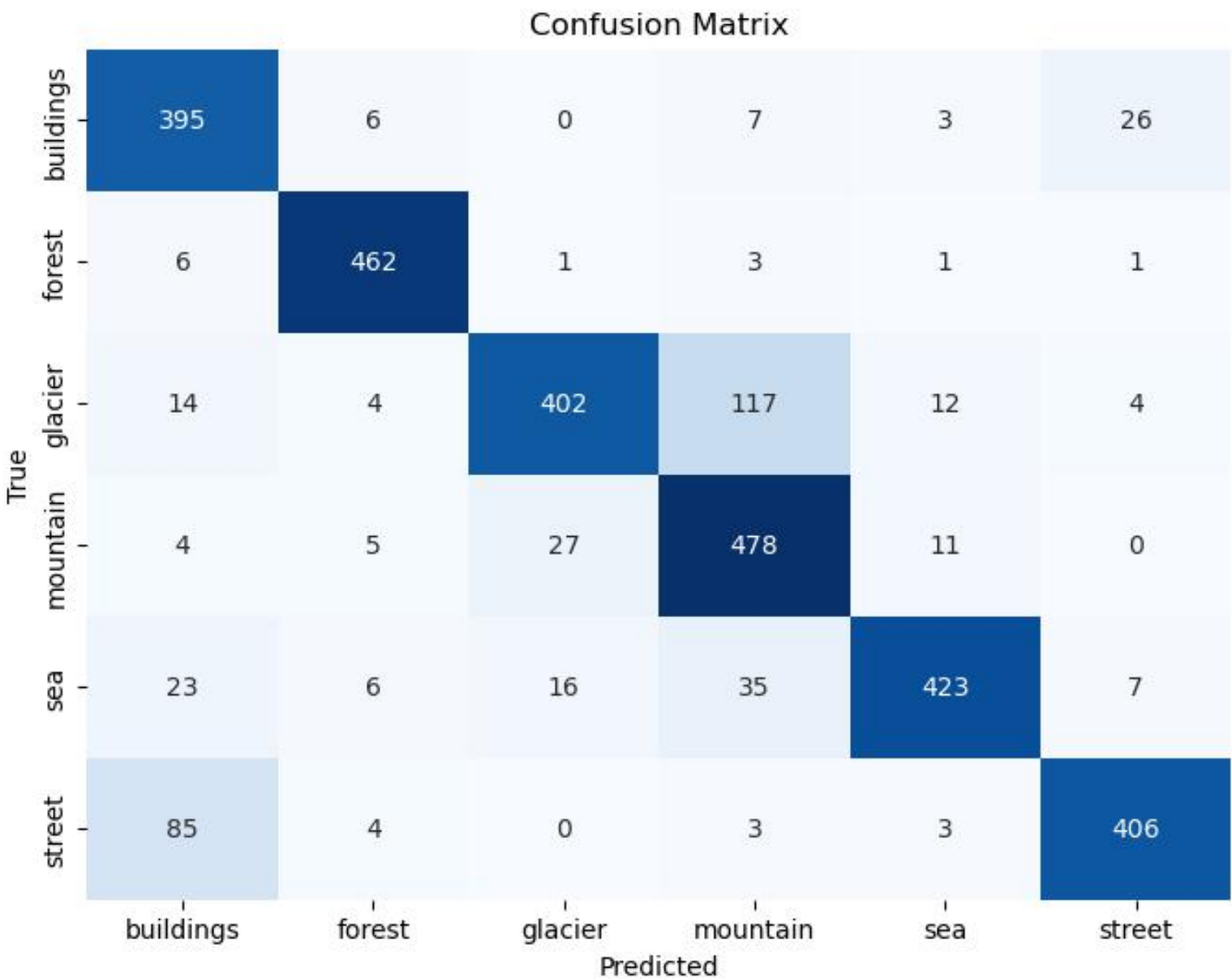


True: sea Predicted: sea



Model 3 (6 Convolutional Layers)

Test: 0.86
Train: 0.87



Mountain or Galcier?



Street or Building?

Best performing model

Baseline

**Fully connected
neural network**

Test: 0.46
Train: 0.43

Model 2

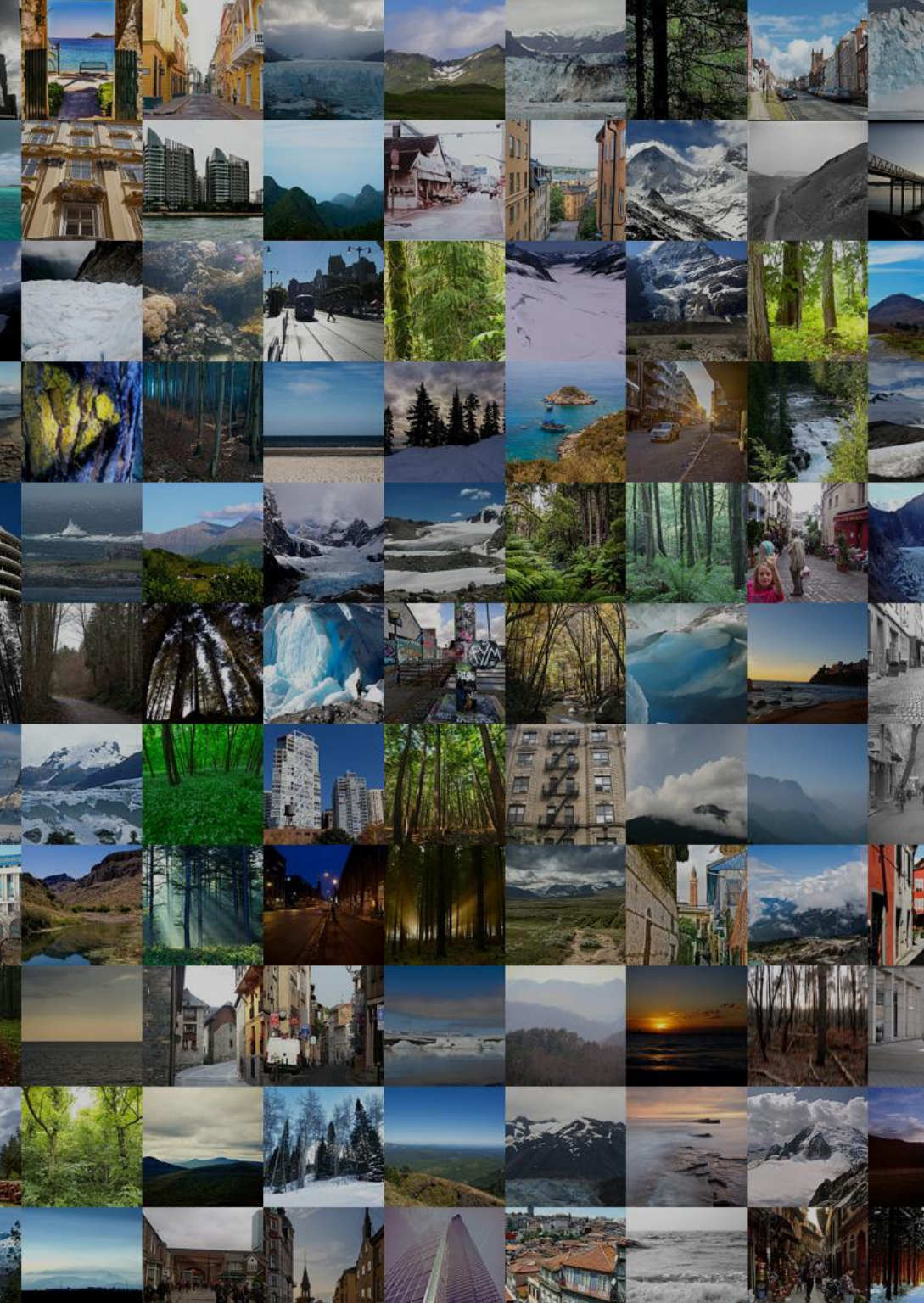
**3 Convolutional
layers**

Test: 0.80
Train: 0.81

Model 3

**6 Convolutional
layers**

Test: 0.86
Train: 0.87



Recommendations

- Begin the implementation of the image classification model for archiving purposes.
- Establish a feedback loop with archivists and users to continually improve the model's performance.
- Explore partnerships and collaboration opportunities with organizations or platforms in related fields, such as content management or digital libraries.

Next steps

create an object detection or scene understanding model that would recognise all the components of the scene even if it included more than one element (Street, building)



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[Linkedin.com/in/alihijazy/](https://www.linkedin.com/in/alihijazy/)

**THANK
YOU**