

# SketchZoo: Animal Image Retrieval with Siamese Neural Networks

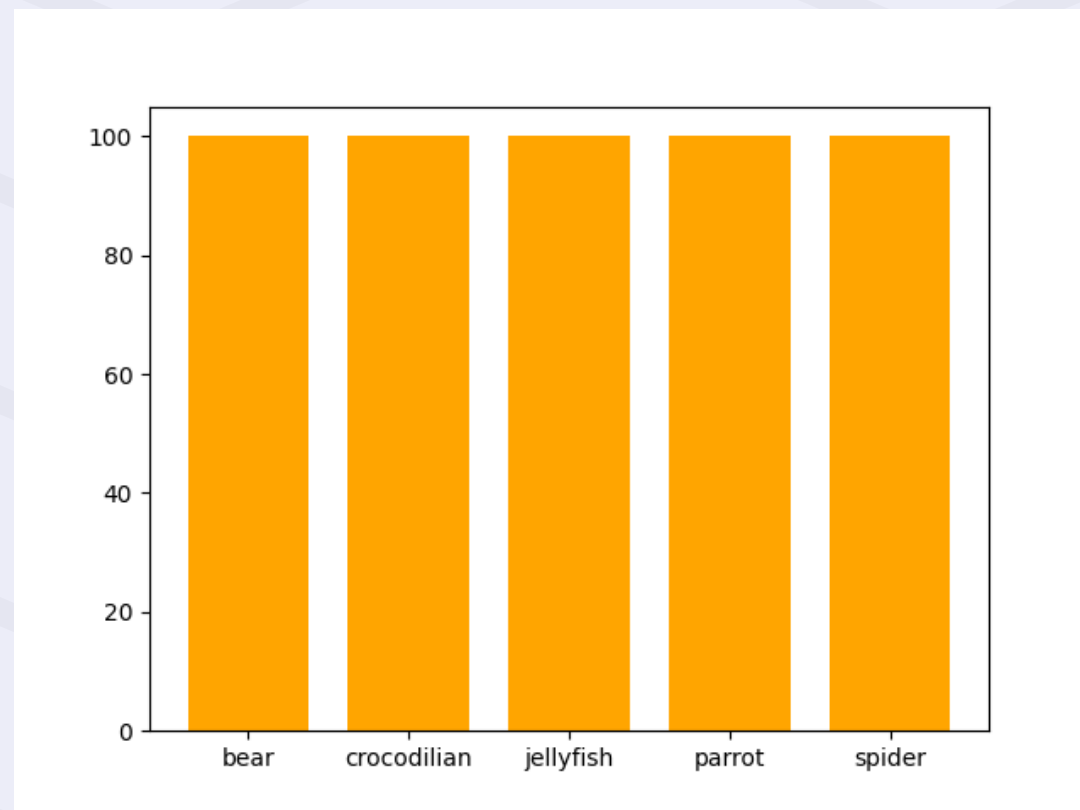
Davide Brescia



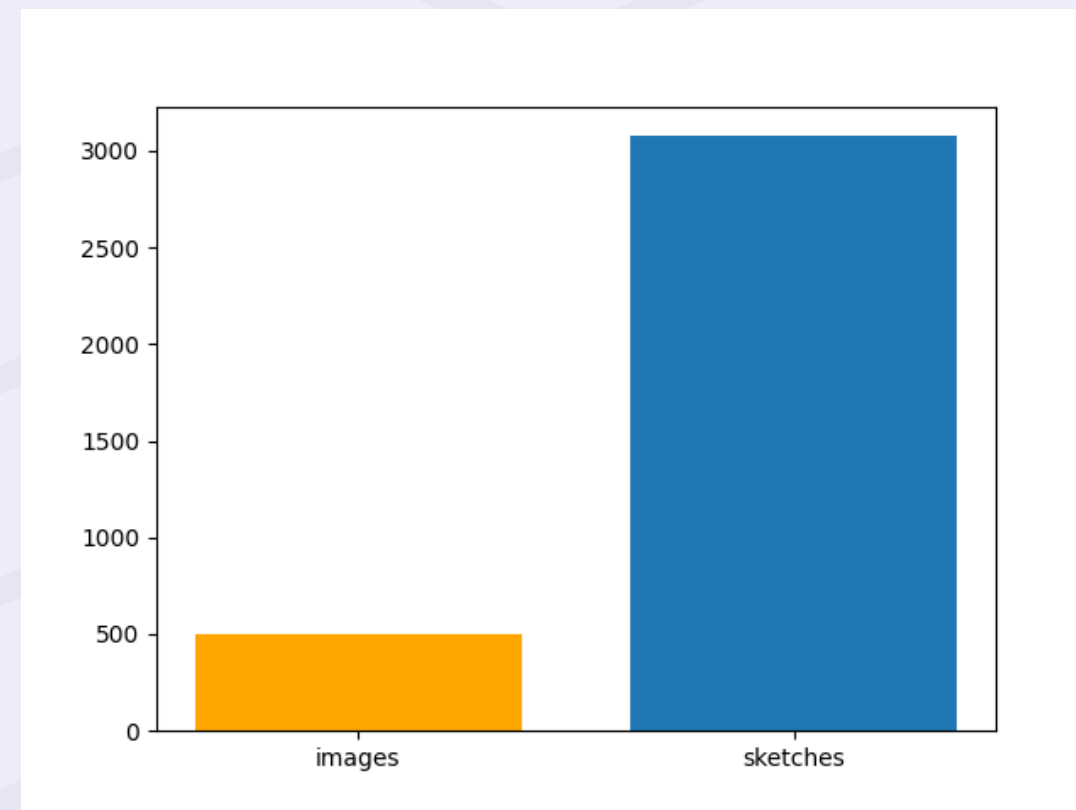
# MINI-DATASET DISTRIBUTION

To ensure the initial functionality of the system, I opted to work with a small subset of the complete dataset. The subset consists of **5 Classes** (bear, crocodilian, jellyfish, parrot, spider) and comprises a total of **500 Images** and **3076 Sketches**.

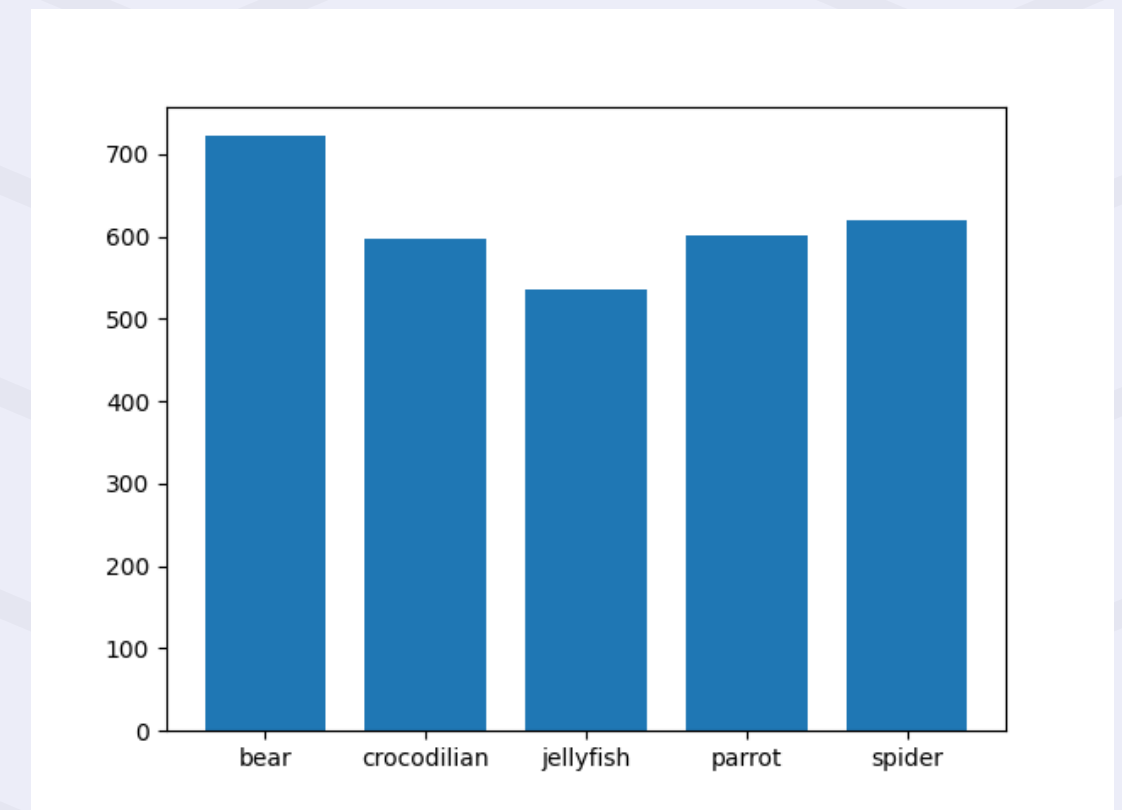
## Images Distributions



## Images vs Sketches



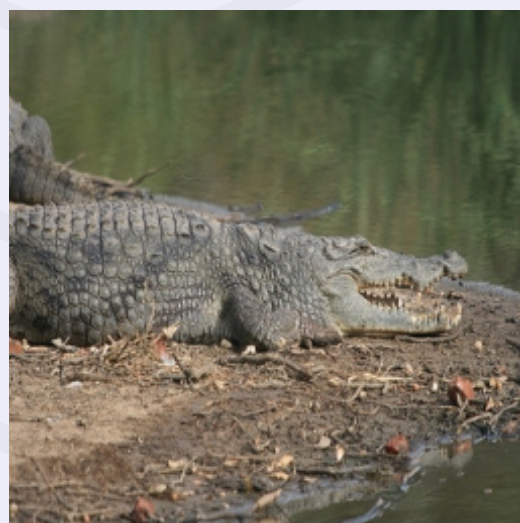
## Sketches Distribution



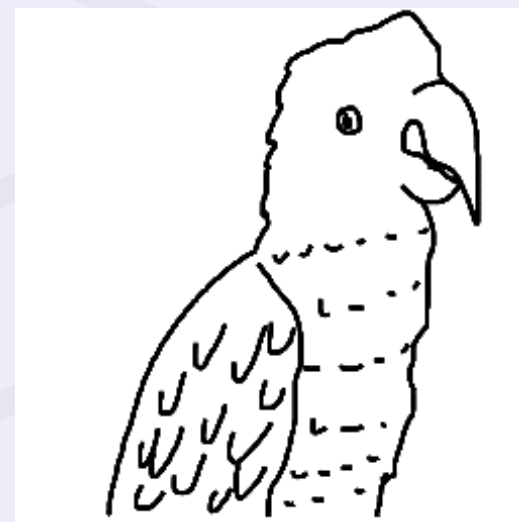
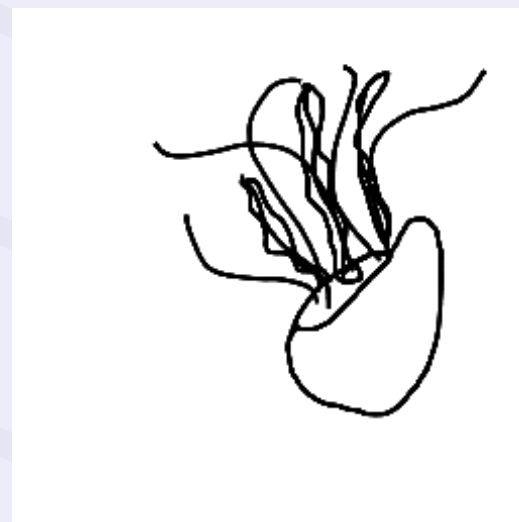
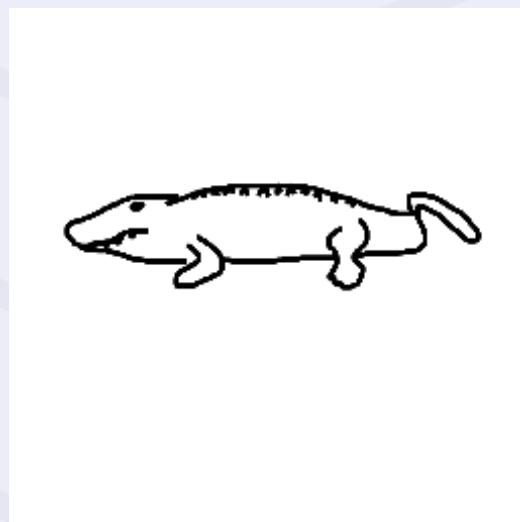


# MINI-DATASET EXPLORATION

## - Some Images -

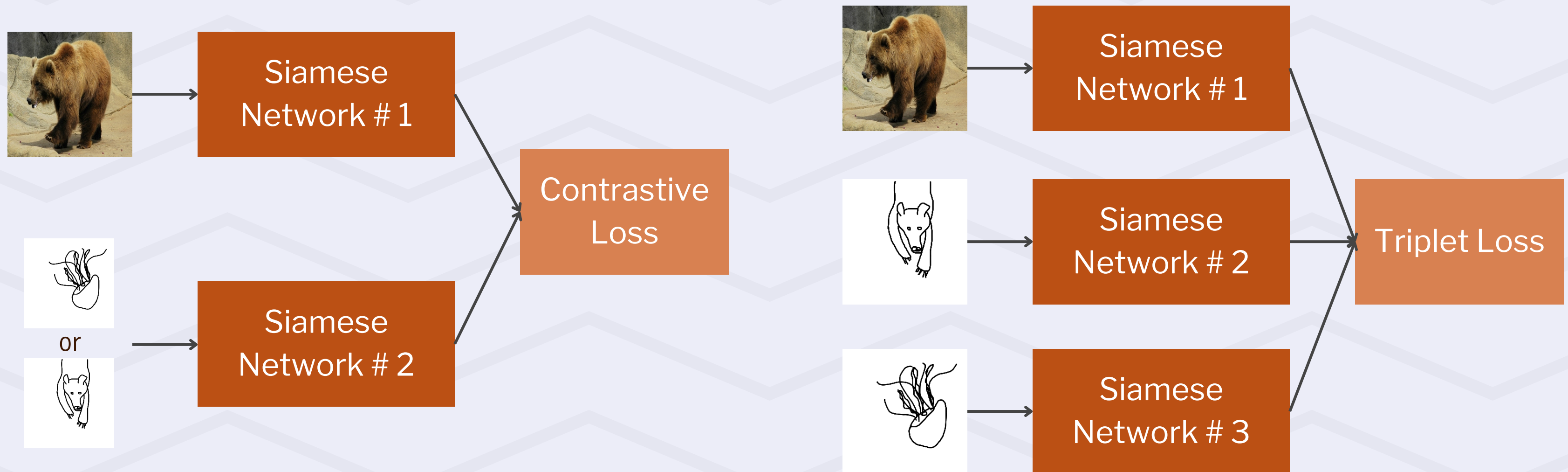


## - Some Sketches -



# PROBLEM APPROACH

To address this issue, the **Siamese Neural Network** was employed, which involves the use of **two** (in the case of contrastive loss) **or three** (in the case of triplet loss) **neural networks** with shared weights. The embeddings are then computed and compared to obtain the loss.



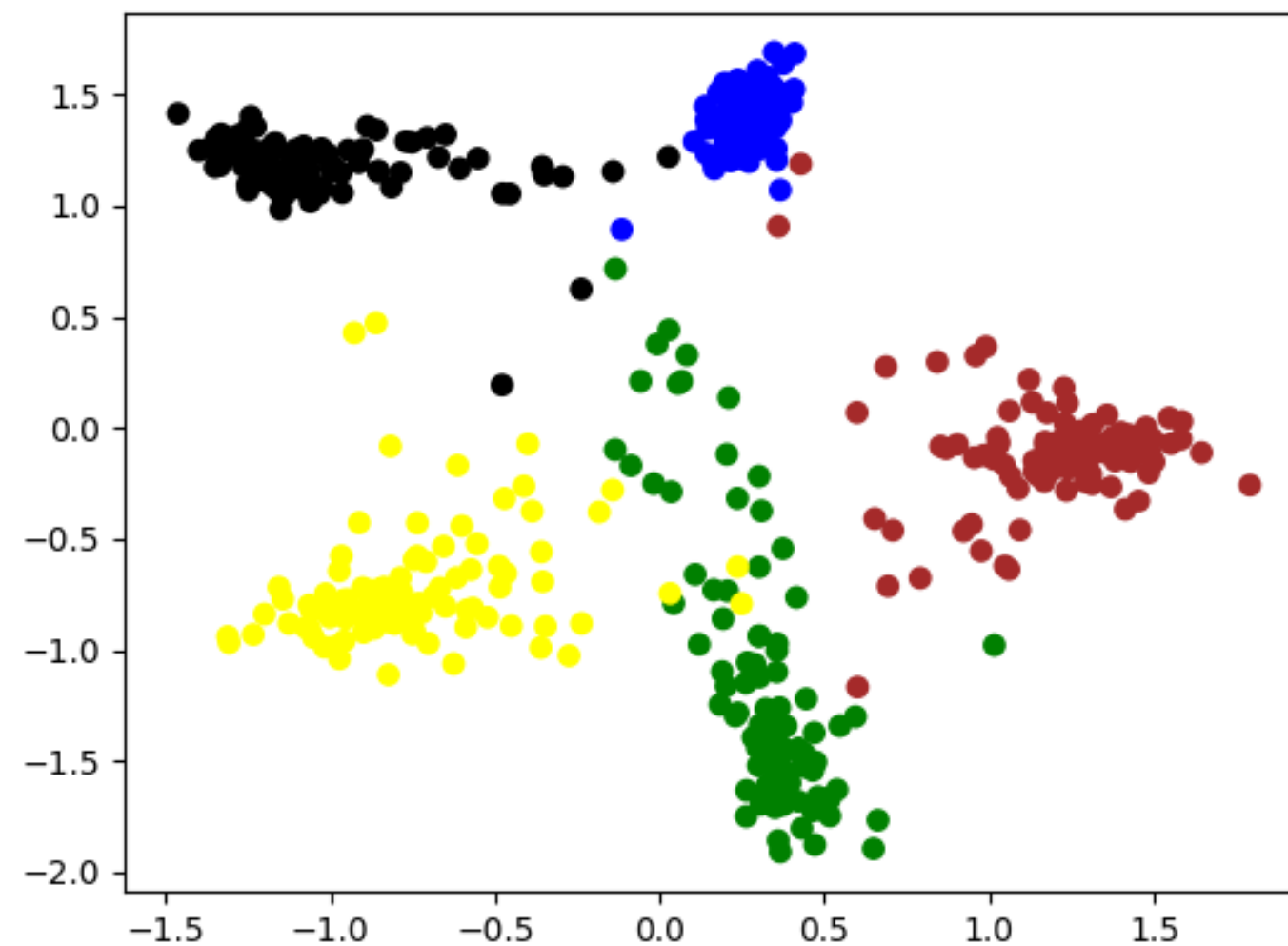
# EXPERIMENT 1 – DETAILS

**Name:** Experiment 1

**Backbone:** ResNet18

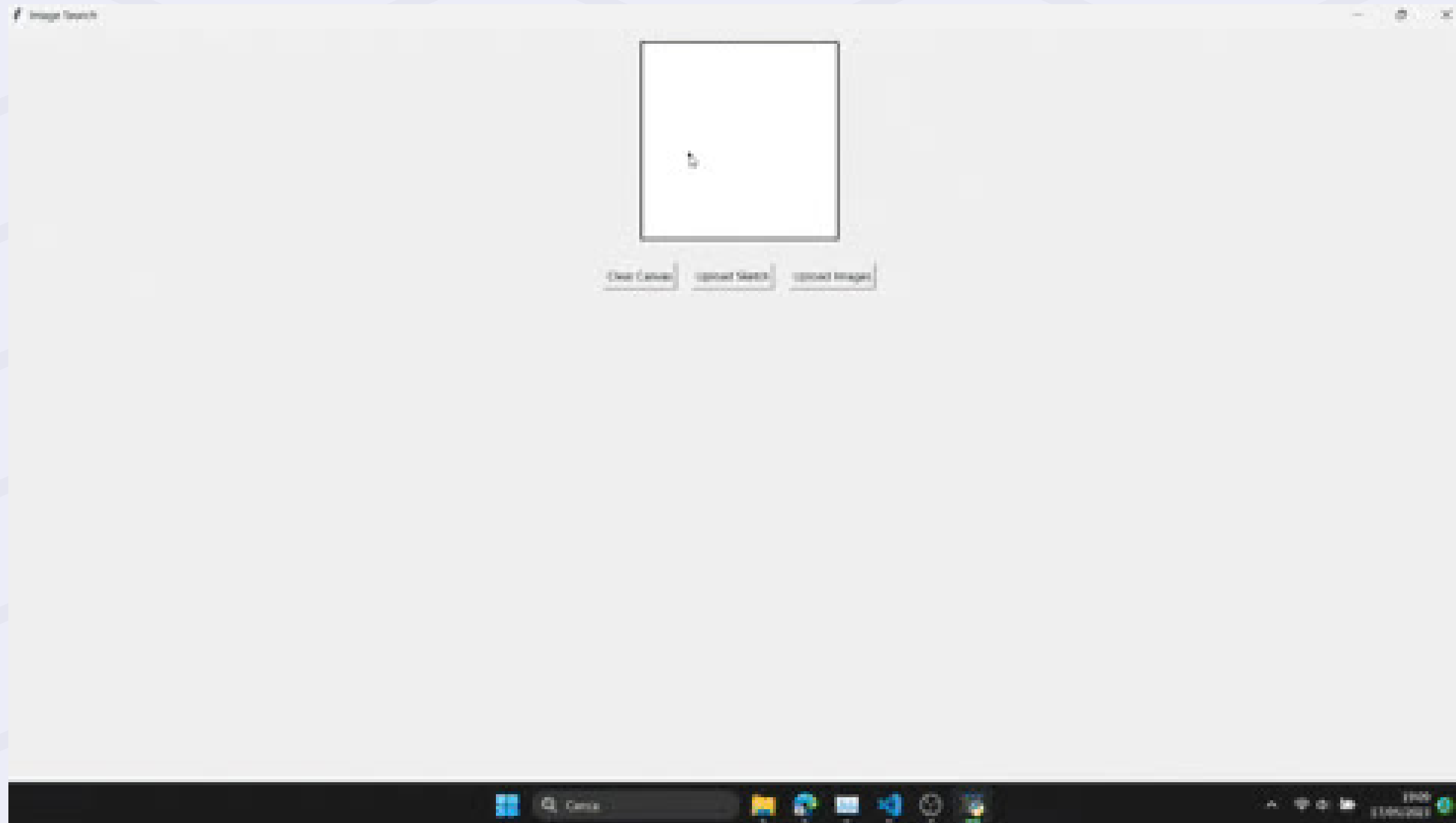
**Loss:** Constrastive Loss

**Embedding Size:** 2



# EXPERIMENT 1 – RESULTS

As we can see the model works and has some ... **interesting results!**



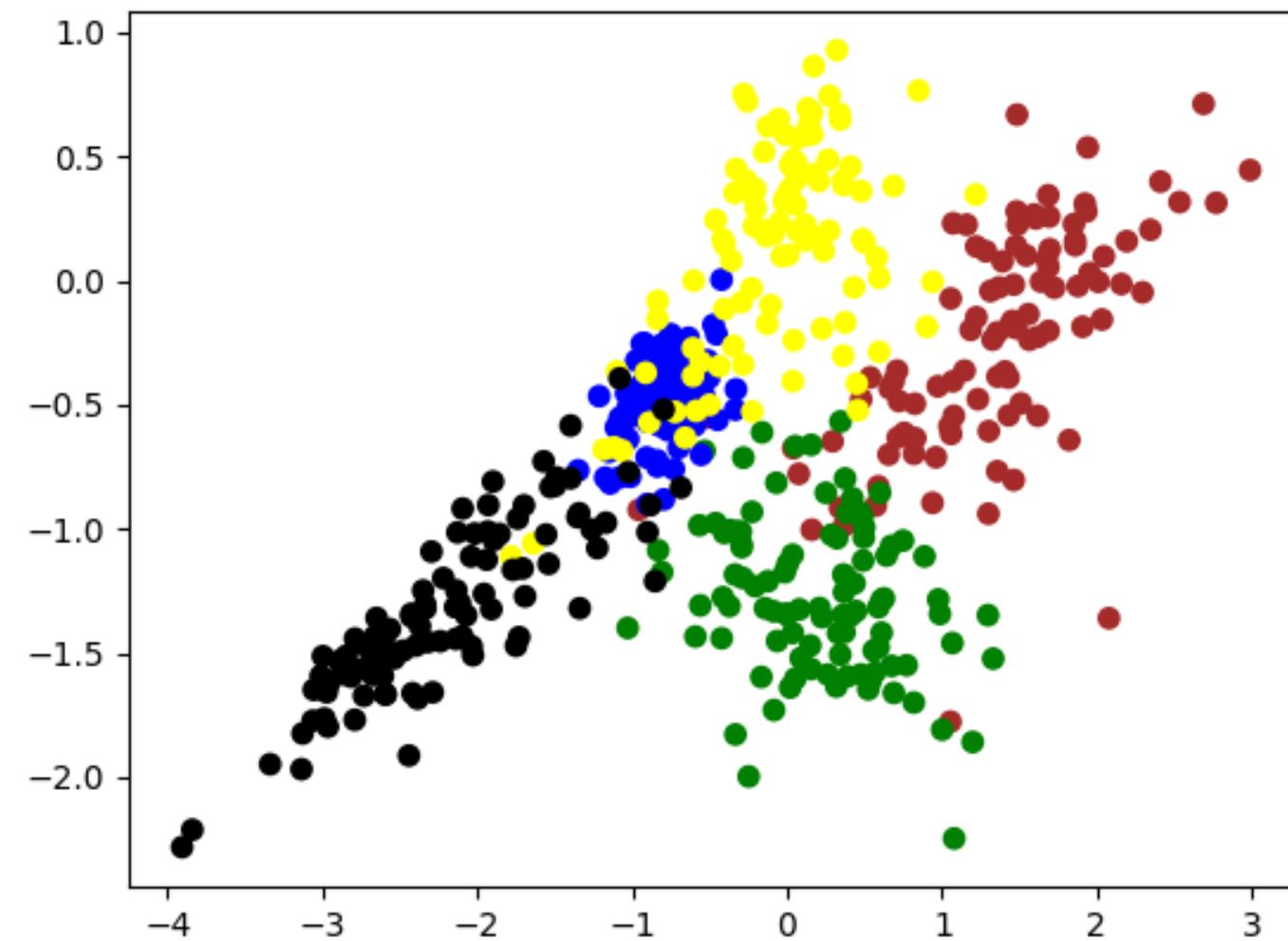
# EXPERIMENT 2 – DETAILS

**Name:** Experiment 2

**Backbone:** ResNet18

**Loss:** Triplet Loss

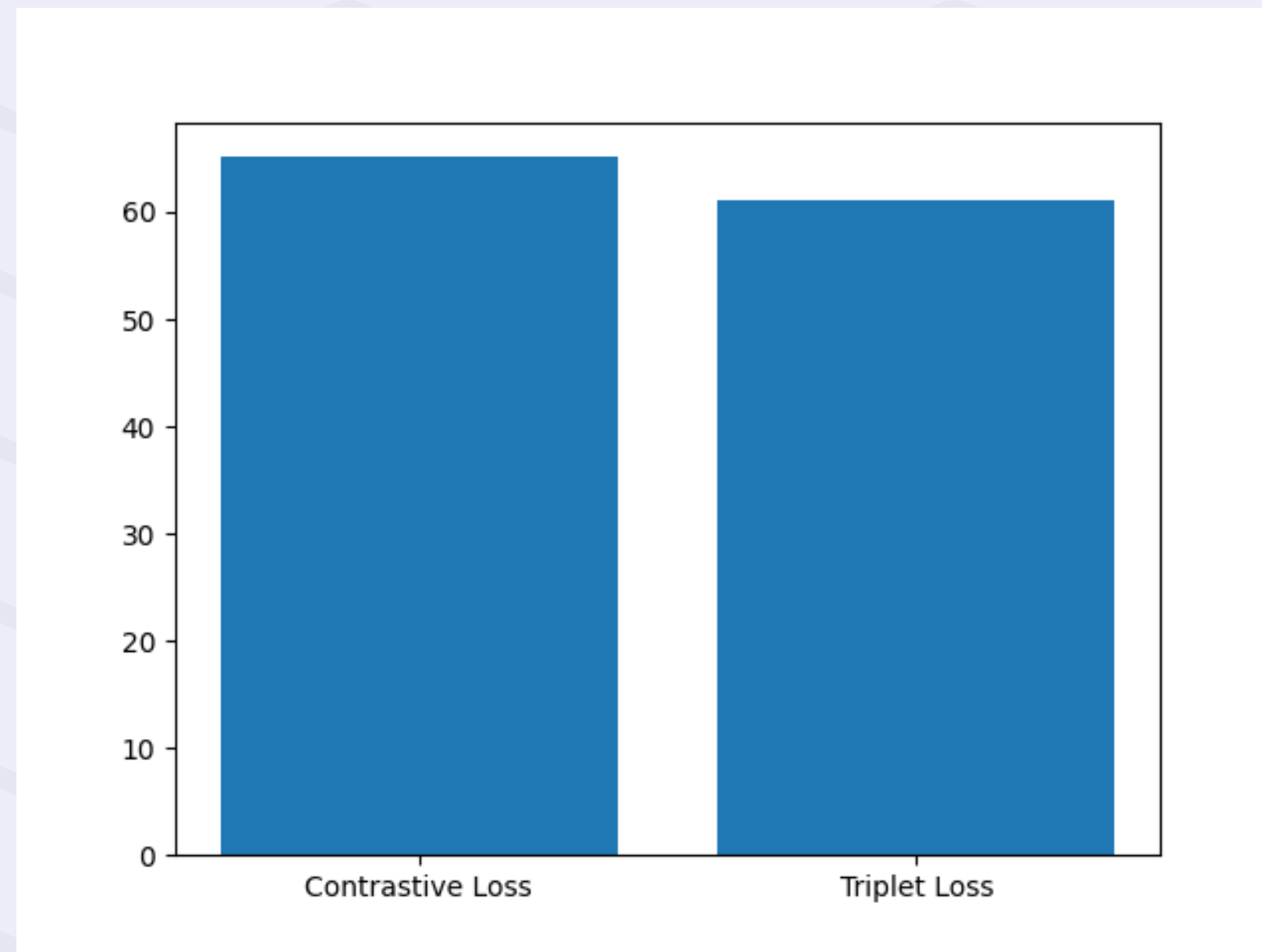
**Embedding Size:** 2





# TRIPLET LOSS VS CONTRASTIVE LOSS

To conduct the evaluation, the **K-Precision** or **P@K** metric was employed, which quantifies the number of accurately identified classes among a set of K images (typically K=12) when applied to unseen images and sketches.

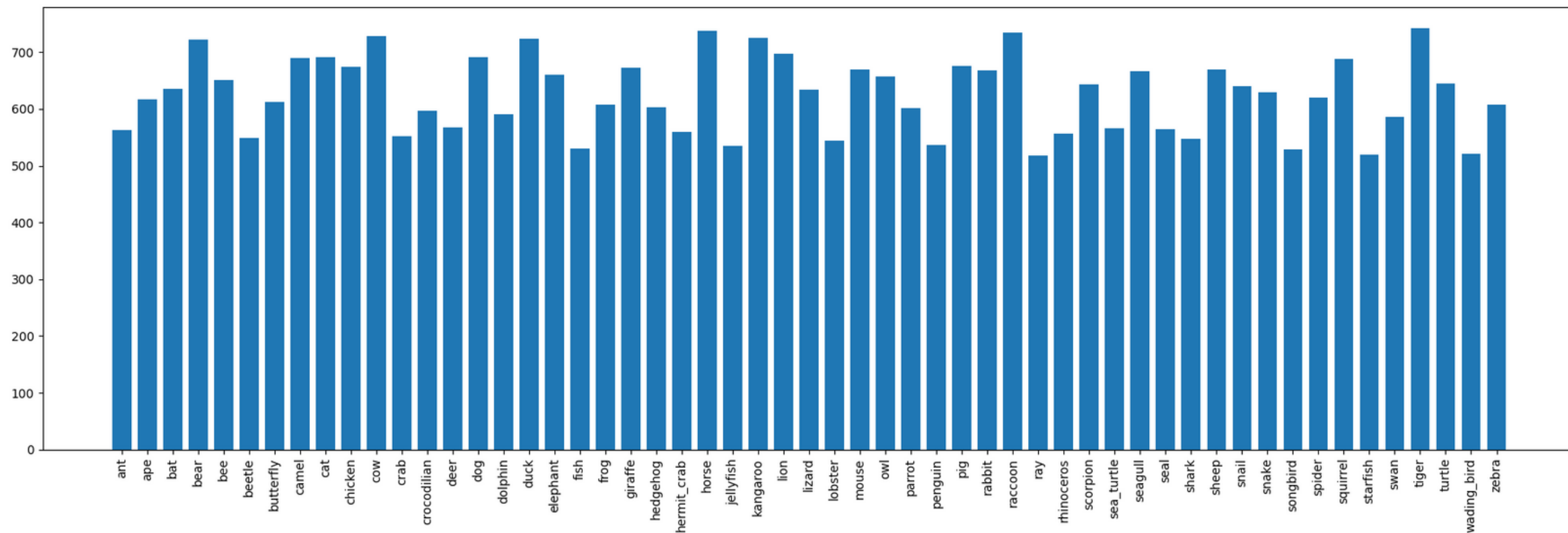




# FULL-DATASET – DETAILS

It is now time to utilize a **larger subset**. Specifically, **55 classes** belonging to the animal kingdom domain have been selected, with **each class associated with 100 images** and a variable number of sketches, totaling **34,366 sketches**.

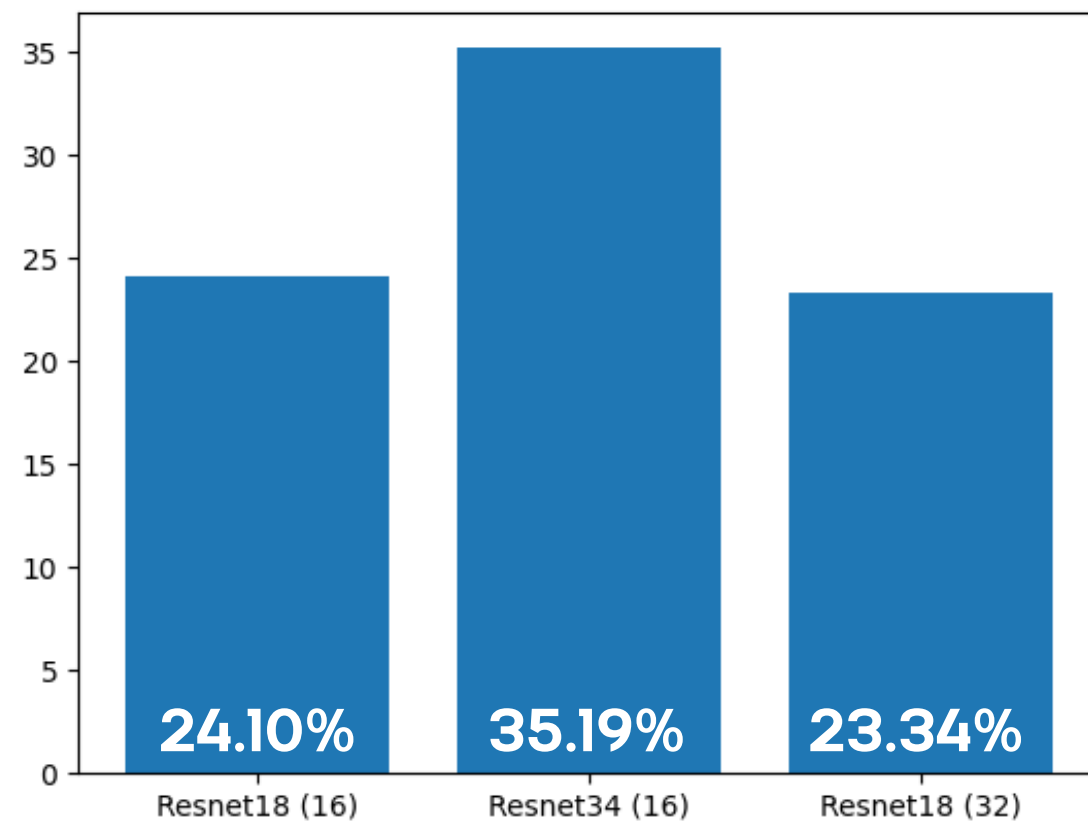
## – Sketches Distribution –



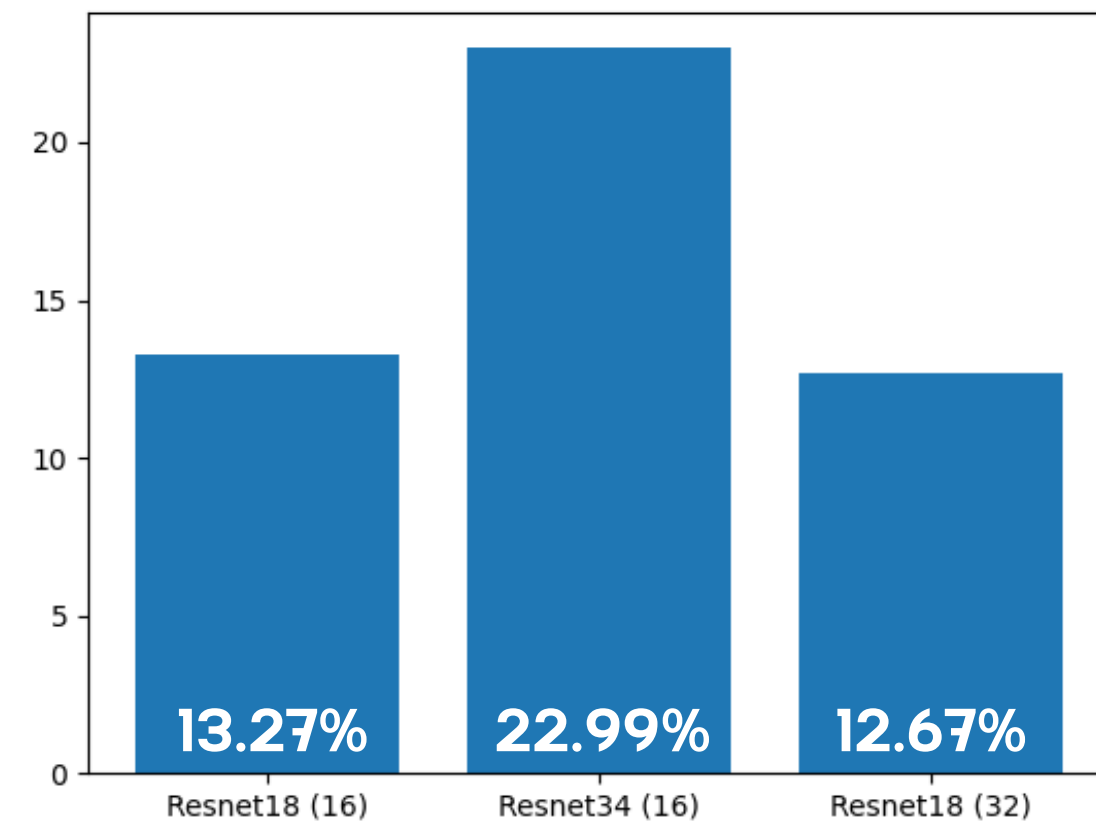
# FULL-DATASET – RESULTS COMPARISON

I conducted additional experiments by using more **complex models** and **increasing the embedding size**. As we can see from the significant increase in the lower graphs, transitioning from **ResNet18** to **ResNet34** yields noticeable improvements.

## – K@12 Contrastive Loss –



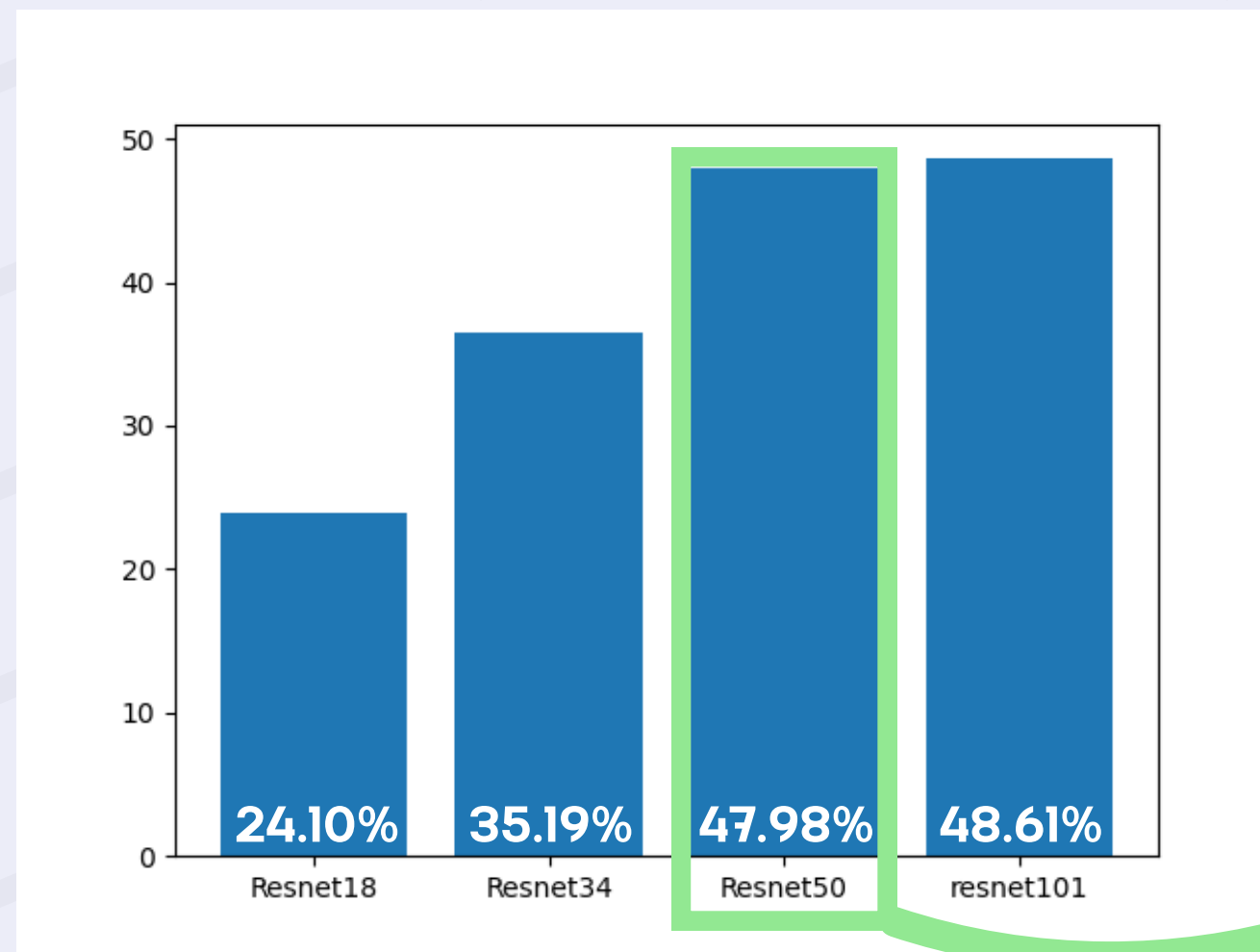
## – K@12 Triplet Loss –



# FULL-DATASET – MODEL IMPROVEMENT

Since the **Contrastive Loss** was leading to better results, I decided to focus solely on that and train **ResNet50** and **Resnet101** as well to see if the achieved results continued to improve. **Not surprisingly, the results are better.**

## – Models Comparison in Contrastive Loss –



In the end, I chose to implement **this model**, which, despite having a lower K@12, has a significantly **shorter inference time.**

# LIVE DEMO



**THANKS FOR  
YOUR ATTENTION**

Davide Brescia