

Crowd Counting

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Introduction:

- Source Data:

Images collected from a webcam in a mall

- Problem statement:

Object detection problem that counts the number of people in each image.

- Project Design:

TensorFlow framework

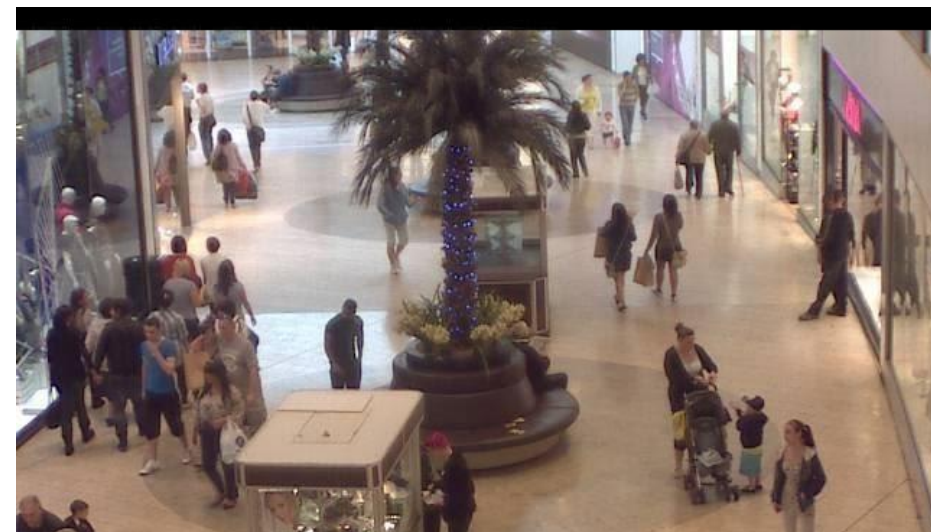
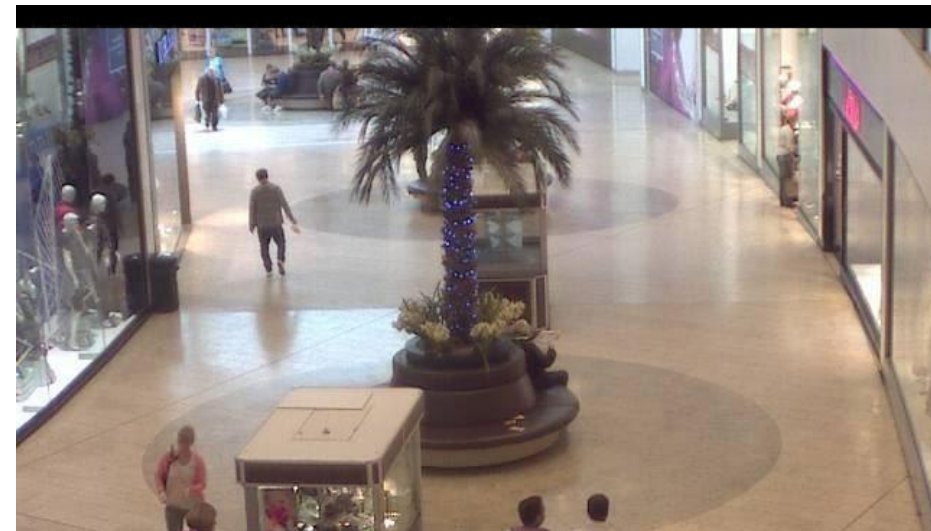
Comparing CNN model with a fine tuned ResNet 50 model

Dataset:

2000 Images

Size:480x640x3

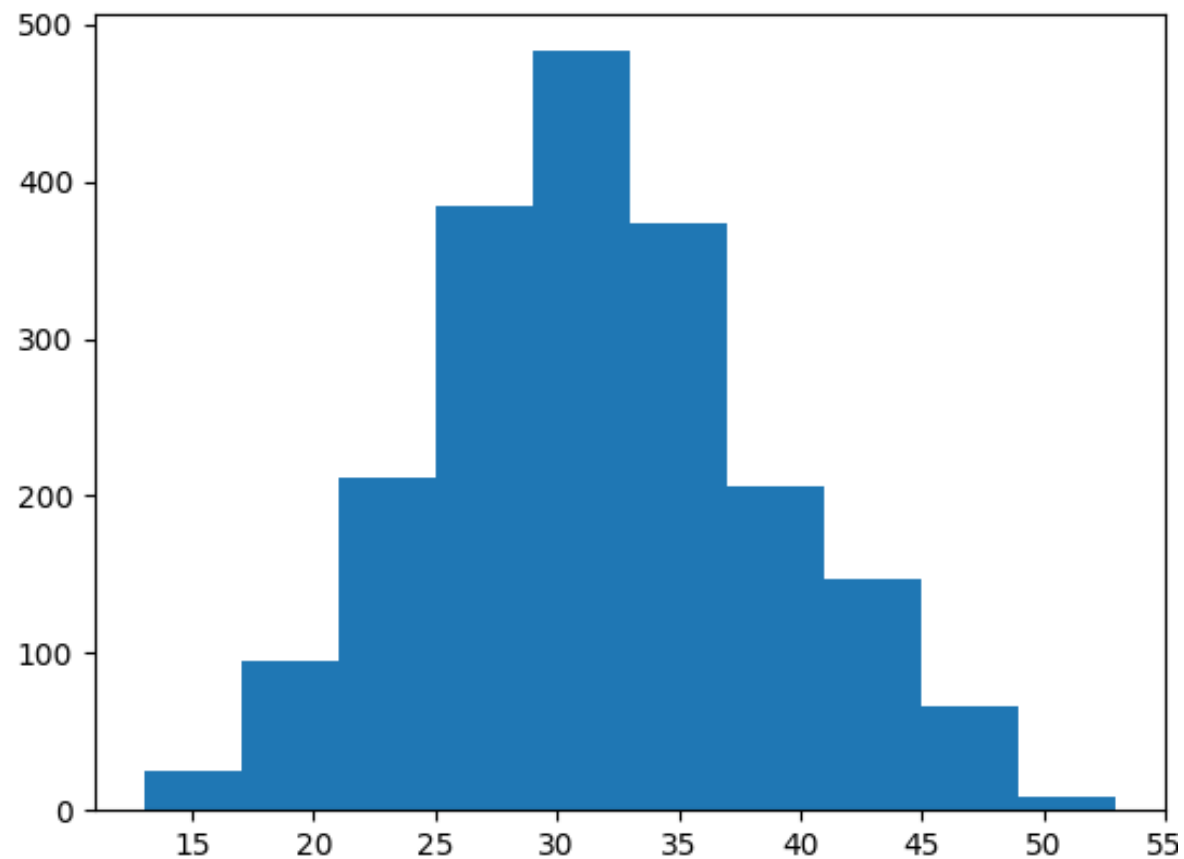
Rezised:128x96x3






Labels:

The data is skewed






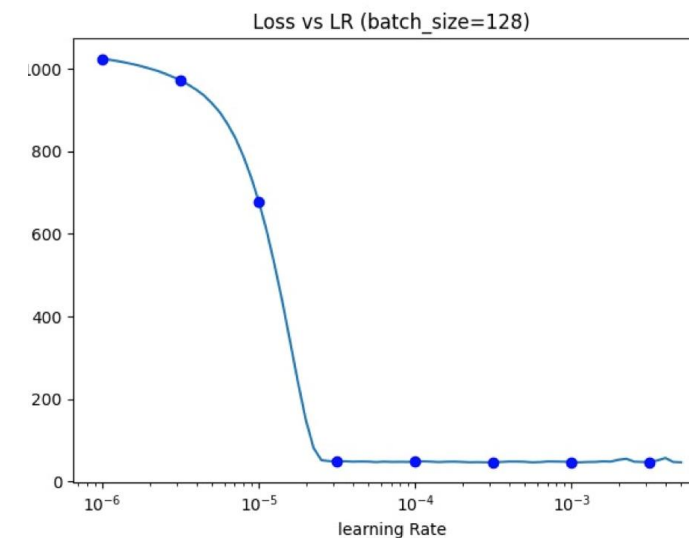
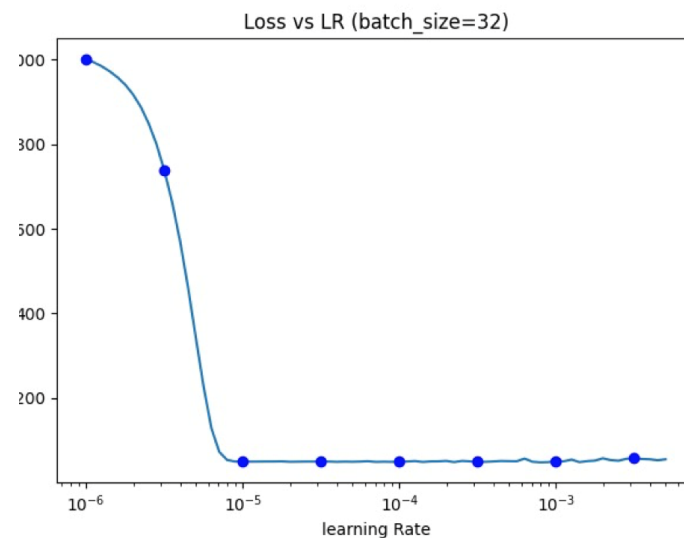
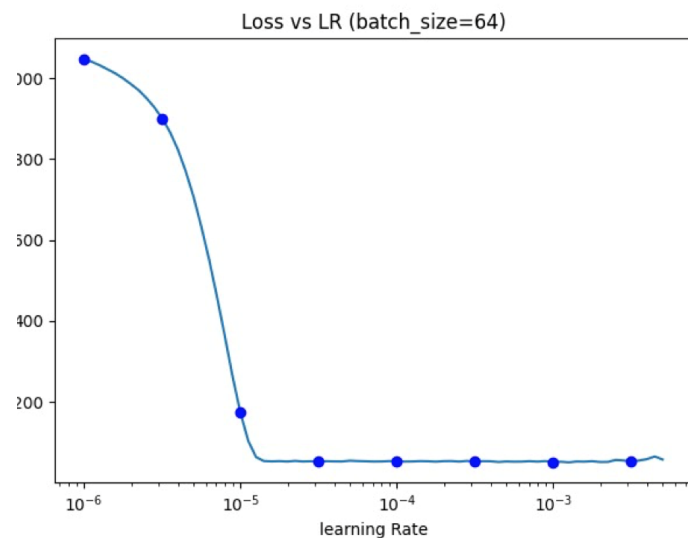
Data Generation and Augmentation

```
53
54 # add ImageDataGenerator
55 datagen = ImageDataGenerator(
56     featurewise_center=False,
57     samplewise_center=False,
58     featurewise_std_normalization=False,
59     samplewise_std_normalization=False,
60     zca_whitening=False,
61     rotation_range=30,
62     width_shift_range=0.1,
63     height_shift_range=0.1,
64     horizontal_flip=True,
65     vertical_flip=False,
66     shear_range=0.5)
67 datagen.fit(x_train)
```

```
61
62 # add a learning rate monitor to get the plot of loss with different learning rate
63 LR_monitor = tf.keras.callbacks.LearningRateScheduler(
64     lambda epochs: 1e-6 * 10 ** (epochs / 20))
65
66 history = model.fit(x_train, y_train, validation_data=(x_test, y_test),
67                     epochs=75, batch_size=32, callbacks=[LR_monitor])
68
```



Hyper
parameter
selection-
Learning
rate



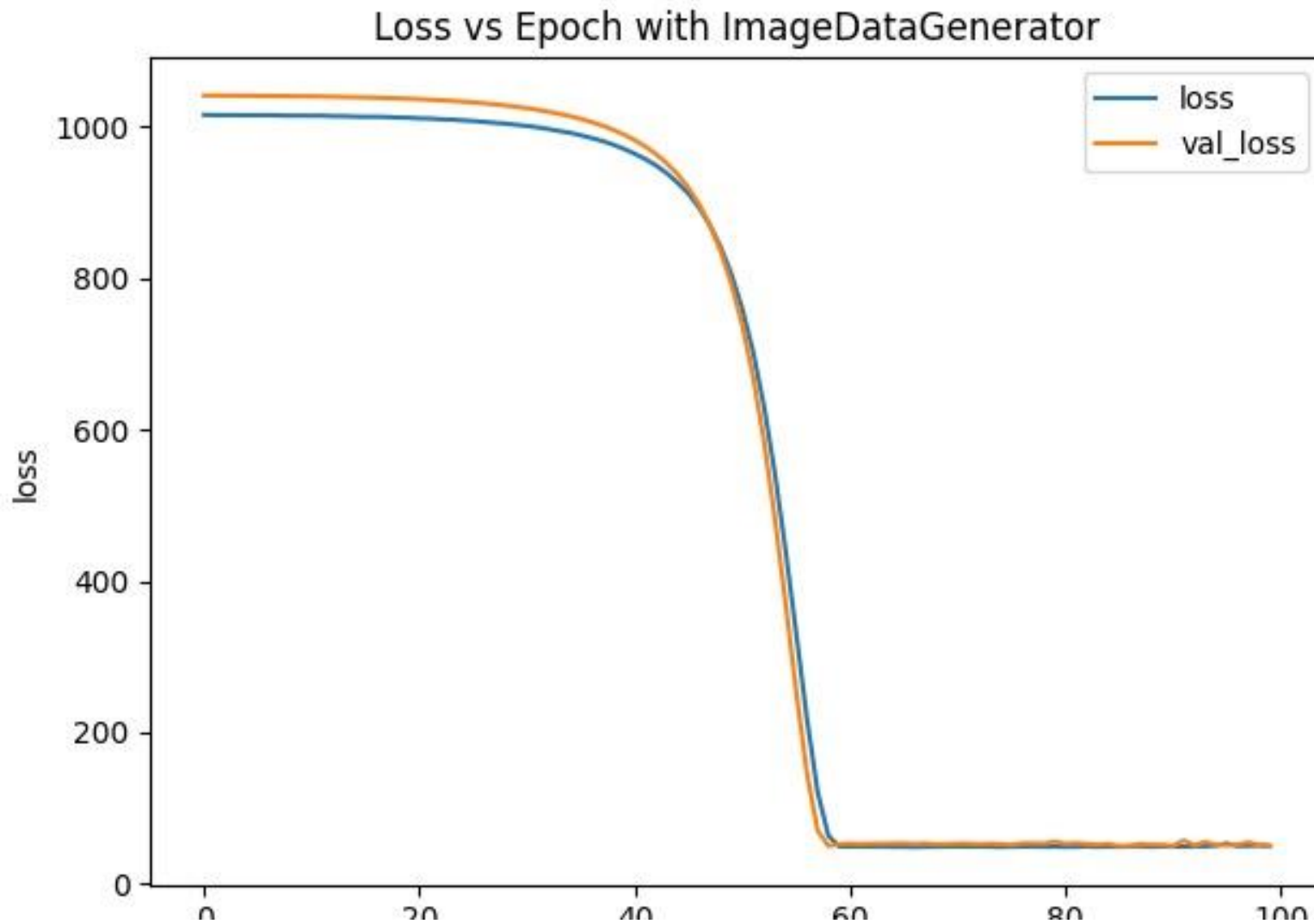
Selecting batch size:

The batch size of 32 is selected

CNN Model

```
2 # create CNN model
3 model = tf.keras.Sequential([
4
5     tf.keras.layers.Conv2D(16, (5, 5), input_shape=(128, 96, 3), activation=tf.keras.activations.relu),
6     tf.keras.layers.MaxPool2D(2, 2),
7     tf.keras.layers.Conv2D(32, (3, 3), activation=tf.keras.activations.relu),
8     tf.keras.layers.MaxPool2D(2, 2),
9     tf.keras.layers.Dropout(0.2),
10    tf.keras.layers.Flatten(),
11    tf.keras.layers.Dense(1)
```

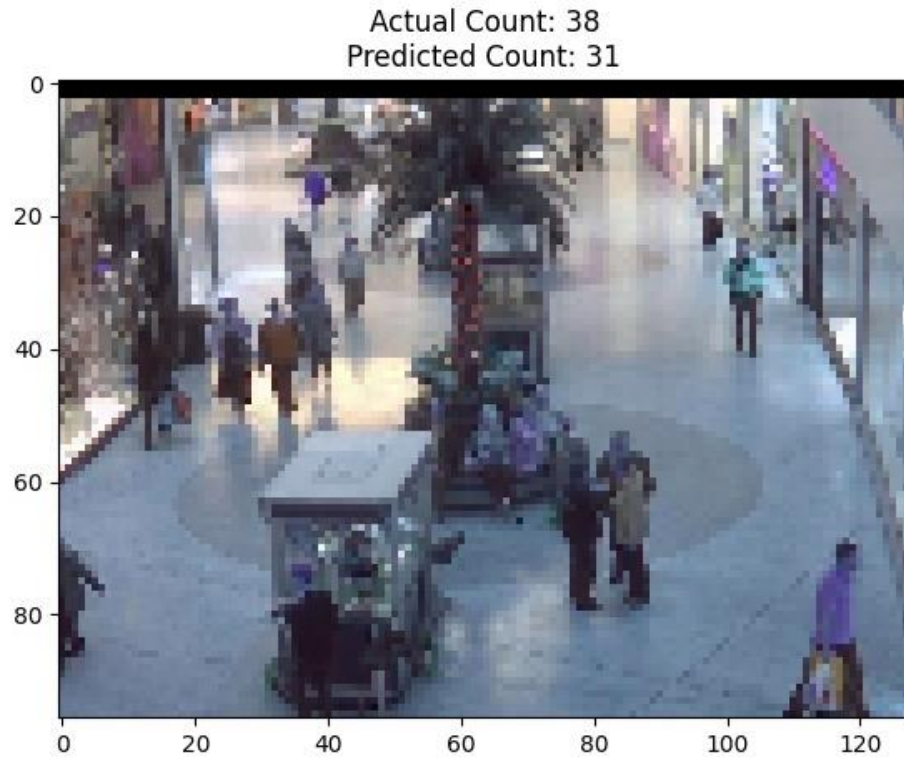
- epochs = 100
- Adam optimizer
- model_check = ModelCheckpoint("keras-CNN-128-96.hdf5",
monitor="val_loss", verbose=1, save_best_only=True)



CNN Model
Evaluation:

val_loss:
49.9336

val_mae:
5.4904



CNN Model Predict Result

ResNet 50:

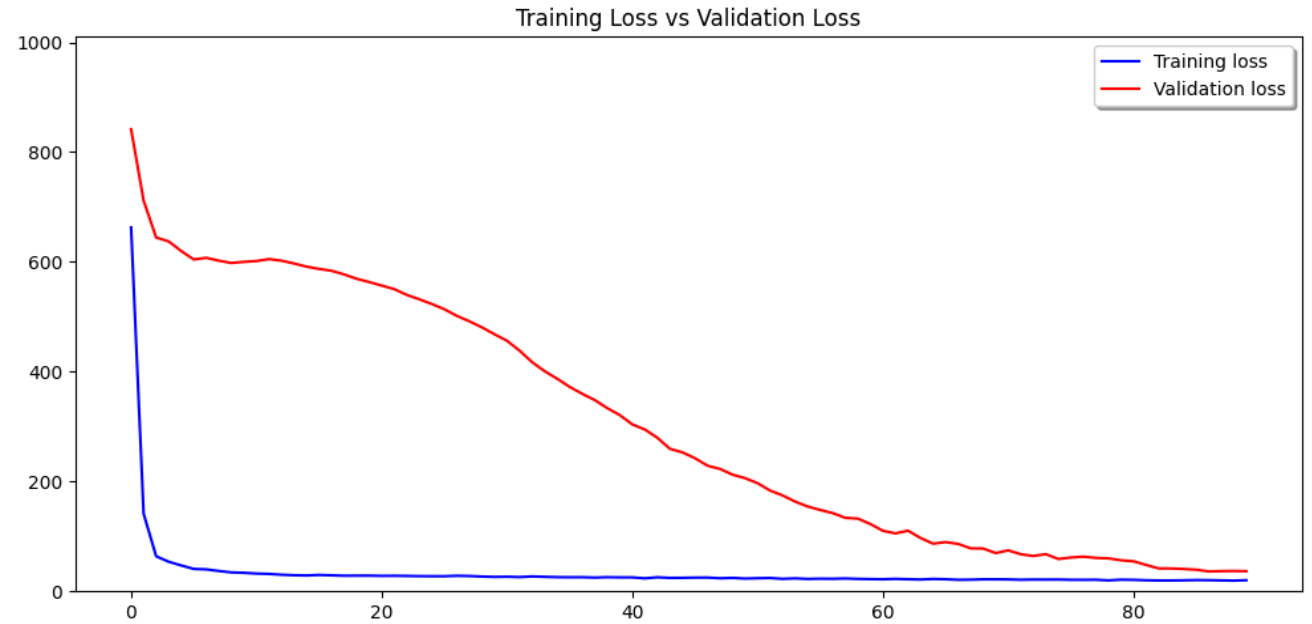
- ResNet-50 is a convolutional neural network that is 50 layers deep.
- 90 Epochs were used for training
- Made last 7 layers untrainable
 - Modify the output layer to predict the number of people in the image
 - Prevent Overfitting
- Adam optimizer used for training
- ReduceLROnPlateau()
 - Monitor on the validation mean square error
 - Learning rate drop 20% every 3 epochs, if the MAE does not go down
 - Lowest learning rate will not go beyond 1e-6

```
learning_rate_reduction = ReduceLROnPlateau(  
    monitor='val_mean_absolute_error',  
    patience=3,  
    verbose=1,  
    factor=0.2,  
    min_lr=0.000001  
)
```

```
optimizer = Adam(learning_rate=3e-4)

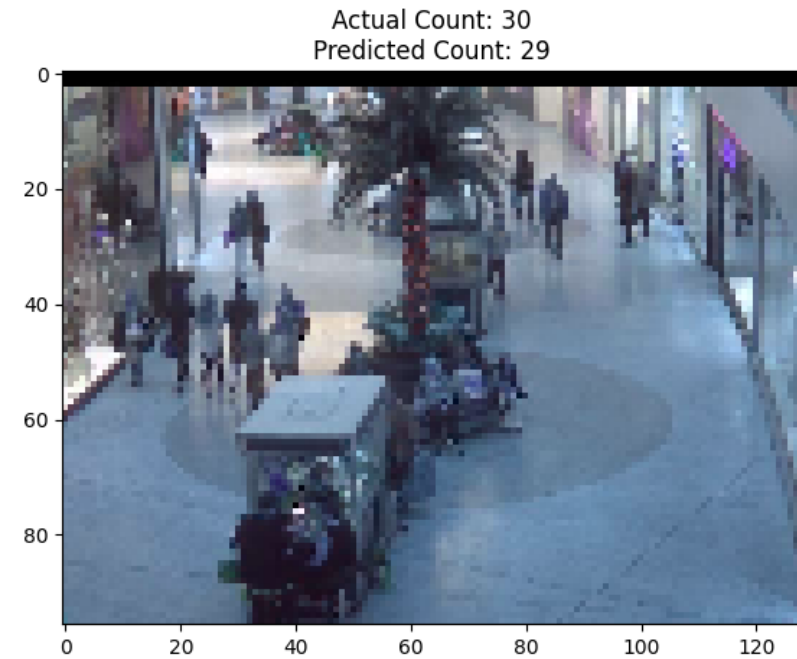
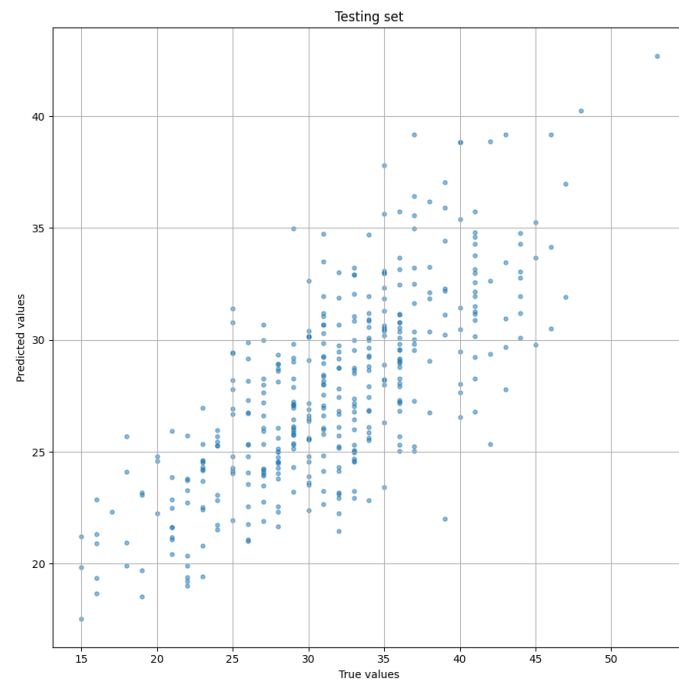
model.compile(
    optimizer=optimizer,
    loss="mean_squared_error",
    metrics=['mean_absolute_error']
)
```

```
MSE Train Set Loss value 20.26082992553711
MSE Validation Set Loss value 36.21799850463867
MAE Train Set Loss value 3.5935373306274414
MAE Validation Set Loss value 4.916306972503662
```



ResNet 50- Results

ResNet 50- Evaluation



Conclusion:

- The ResNet50 model performs better

Models	Mean Absolute Error	Mean Square Error
CNN	5.55	50.29
ResNet 50	4.92	36.31

Future Scope

- MCNN model
- Attempt to decrease loss value by ensembling models
- Find more relevant datasets



Thank you!