#### Capstone 2 Fruit Classifier

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#### Problem

Customers buy fresh fruits in grocery stores. We want to automatically recognize type of the fruit by the scale – fruit type.

#### Data source

The source of data is kaggle fruit-recognition data set.

- Size of data is 8 GB.
- Each image is size (200, 200), (320, 258) or (480, 322).
- Only one image is size (320, 240).
- There are 54.

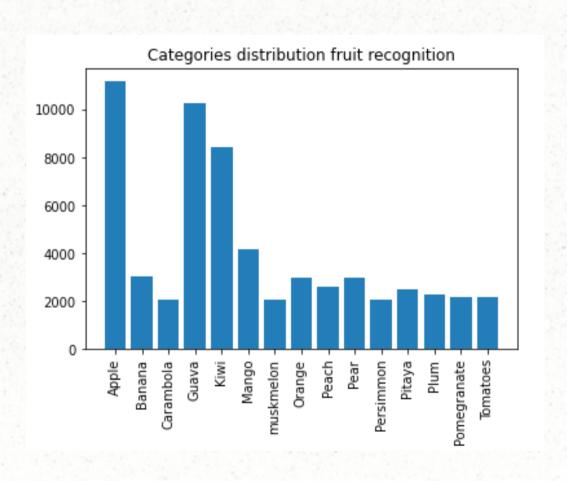
#### Data - classes

The images represent 15 different fruits.

- 3 classes have more than one sub-folder containing images;
- 9 classes have "resized" or "resized resized" images 200 by 200 pixels.

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### Data distribution



### Data findings

We found 15 type of images.

- Class Apples has the most images.
- Classes Guava and Kiwi has similar, but smaller amount images.
- All other classes have about 2000 to 4000 images each.

#### Data

We are going to use the images which are the most in the dataset.

We created file with names of images same size.

Further, we mix all names in train test split and saved for machine learning.

Common architecture used for computer vision is convolutional neural network.

 Common ways to transform image is convolution, maxpooling or filtering.

- Convolution is data manipulation on image pixels. In the core is calculating values of each pixel of a feature map by multiplying intensity value of part of the image and a small tensor.
- It is used to learn local patterns.
- At each level, we apply multiple convolutions.

- Maxpooling is a way to change the scale of image. It is applied between convolutions, so the convolution will learn local patterns at different scale.
- We can also use average of all pixels to change the scale.

There are two main approaches to create an architecture.

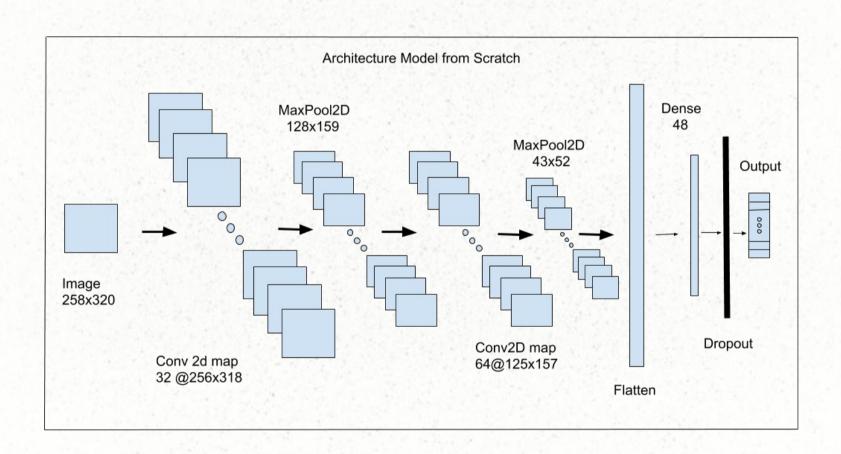
- We can design model from scratch. In this case, we choose all parameters and structure.
- We can use frozen base and attach new classifier to learn the classes.

When re-purposing model, we can use already fitted values and only re-train the last layers.

Other option is to train the whole model from scratch, only using the structure.

Our data set is small, we used the first approach.

# Model from scratch



### Training of the model

Training the model takes three pieces: optimizer, loss function, and metric to measure performance.

- The optimizer is the algorithm to search for values of the connections that will minimize the loss function.
- We used a few optimizer algorithms to fit.

### Training of the model - loss

Training the model takes three pieces: optimizer, loss function, and metric to measure performance.

• The loss function for multi-class problems is called cross-entropy.

### Training of the model - metrics

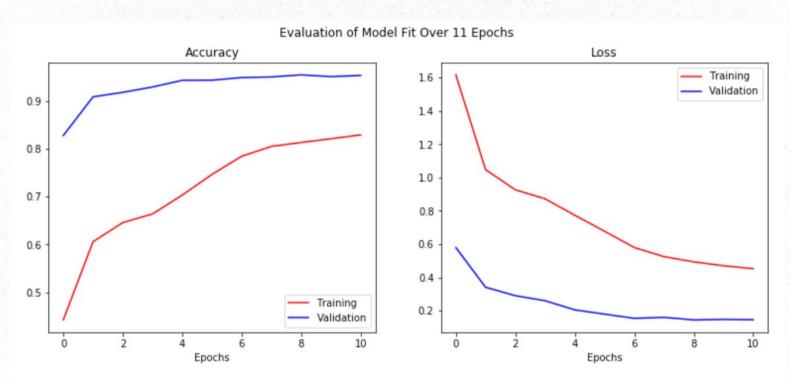
Training the model takes three pieces: optimizer, loss function, and metric to measure performance.

• We measure the accuracy of classifying images.

### Base model performance

- Base model was trained using 11 epochs
- We used early stopping monitor with patience 2.

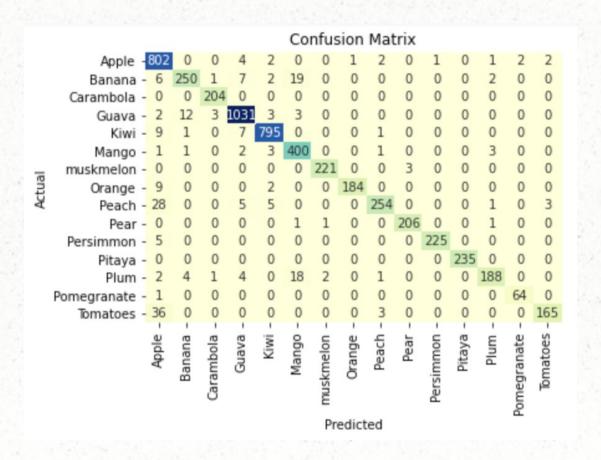
# Base model performance



history\_bsln.csv

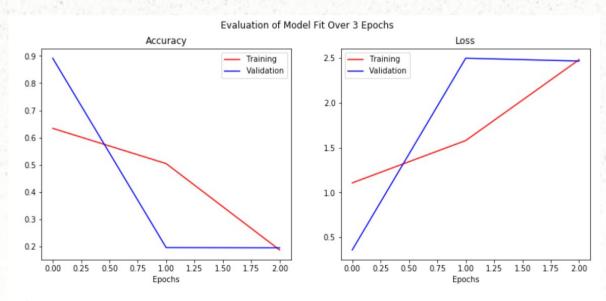
Epochs: 11; Final accuracy: 0.829 Epochs: 11; Final val accuracy: 0.953

#### Base model test



### Other optimizers

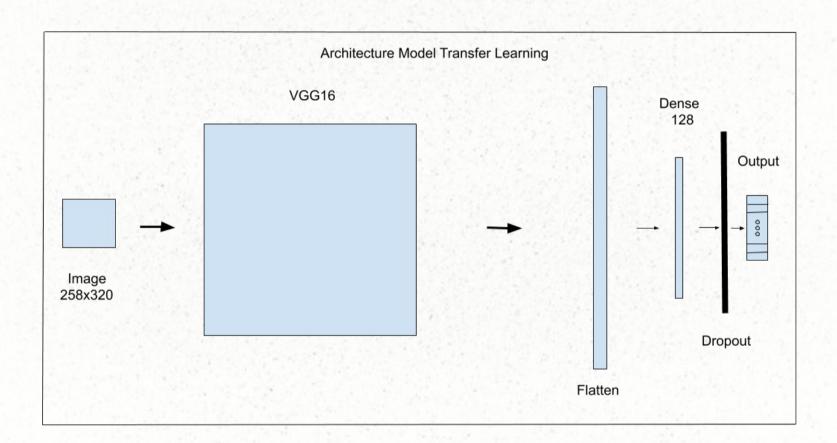
One of the optimizers did not perform well. Possible reasons – large learning rate or wrong batch size.



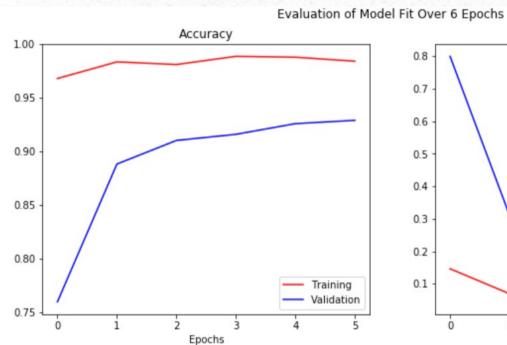
history nadam.csv

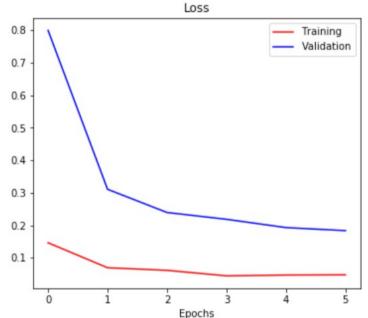
Epochs: 3; Final accuracy: 0.187
Epochs: 3; Final val accuracy: 0.195

# Transfer learning



# Transfer learning training

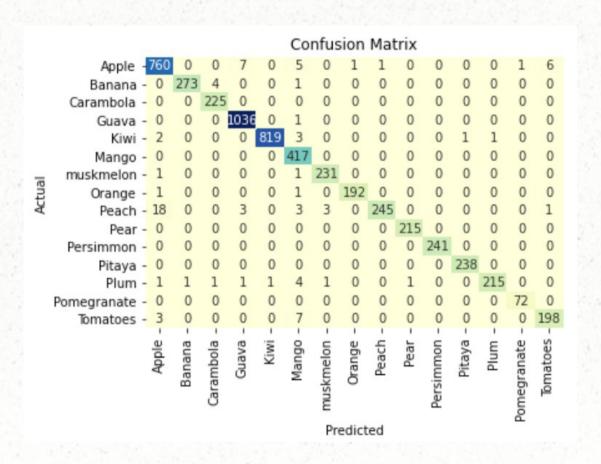




history\_transfer.csv

Epochs: 6; Final accuracy: 0.984 Epochs: 6; Final val accuracy: 0.929

## Transfer learning - test



### Conclusion

Two main approaches were presented here.

- We built classifier from scratch to recognize the name of fruits.
- We used pretrain classifier to recognize the images.

### Conclusion

- The pretrained classifier performed better.
- It has higher accuracy
- It was trained in half of the time, even so it has much more complex architecture.

### Conclusion

The classifier was trained on about 50000 images. We did not use augmentation and rescaling.

More images can be collected.

We also can re-train the model, if more types of images are available.