**Topic:** Al-powered Nutrition Analyzer for Fitness Enthusiasts

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## **MODEL BUILDING**

Here we are going to build our Convolutional Neural Networking which contains an input layer along with the convolution, maxpooling, and finally an output layer.

This is an important step because the model is the main thing needed for prediction.

# • Importing The Model Building Libraries

We have imported all the necessary libraries needed for model building.

## Initializing The Model

Keras has 2 ways to define a neural network:

- Sequential
- Function API

The Sequential class is used to define linear initializations of network layers which then, collectively, constitute a model. We will use the Sequential constructor to create our model, which will then have layers added to it using the add() method.

# Adding CNN Layers

- As the input image contains three channels, we are specifying the input shape as (64,64,3).
- We are adding a two-convolution layer with an activation function as "relu" and with a small filter size (3,3) and the number of filters (32) followed by a max-pooling layer.
- Max pool layer is used to downsample the input (Max pooling is a pooling operation that selects the maximum element from the region of the feature map covered by the filter)
- Flatten layer flattens the input. Does not affect the batch size.

# ADDING CNN LAYERS Nutrition Image Analysis using CNN [1] from google.colab import drive drive.mount('/content/drive/') Mounted at /content/drive/ [2] from google.colab import files uploaded = files.upload() Choose Files 0\_100.jpg • 0\_100.jpg(image/jpeg) - 4773 bytes, last modified: 6/9/2020 - 100% done Saving 0\_100.jpg to 0\_100.jpg

# Adding Dense Layers

- A dense layer is a deeply connected neural network layer.
   It is the most common and frequently used layer.
- The number of neurons in the Dense layer is the same as the number of classes in the training set. The neurons in the last Dense layer, use softmax activation to convert their outputs into respective probabilities.
- Understanding the model is a very important phase to properly use it for training and prediction purposes. Keras provides a simple method, and a summary to get the full information about the model and its layers.

# ADDING DENSE LAYERS MODEL SUMMARY [19] # Initializing the CNN classifier = Sequential() # First convolution layer and pooling classifier.add(Conv2D(32, (3, 3), input\_shape=(64, 64, 3), activation='relu')) classifier.add(MaxPooling2D(pool\_size=(2, 2))) # Second convolution layer and pooling classifier.add(Conv2D(32, (3, 3), activation='relu')) # input shape is going to be the pooled feature maps from the previous convolution layer classifier.add(MaxPooling2D(pool\_size=(2, 2))) # Flattening the layers classifier.add(Flatten()) # Adding a fully connected layer classifier.add(Dense(units=128, activation='relu')) classifier.add(Dense(units=5, activation='softmax')) # softmax for more than 2

# Configure The Learning Process

- The compilation is the final step in creating a model. Once the compilation is done, we can move on to the training phase. The loss function is used to find errors or deviations in the learning process. Keras requires a loss function during the model compilation process.
- Optimization is an important process that optimizes the input weights by comparing the prediction and the loss function.
   Here we are using adam optimizer.
- Metrics are used to evaluate the performance of your model. It is similar to the loss function, but not used in the training process.

# CONFIGURE THE LEARNING Compiling the model [21] # Compiling the CNN # categorical\_crossentropy for more than 2 classifier.compile(optimizer='adam', loss='sparse\_categorical\_crossentropy', metrics=['accuracy'])

## Train The Model

Now, let us train our model with our image dataset. The model is trained for 15 epochs and after every epoch, the current model state is saved if the model has the least loss encountered till that time. We can see that the training loss decreases in almost every epoch till 15 epochs and probably there is further scope to improve the model.

fit\_generator functions used to train a deep-learning neural network Arguments:

- steps\_per\_epoch: it specifies the total number of steps taken from the generator as soon as one epoch is finished and the next epoch has started. We can calculate the value of steps\_per\_epoch as the total number of samples in your dataset divided by the batch size.
- Epochs: an integer and number of epochs we want to train our model for.
- validation\_data can be either:
  - an inputs and targets list
  - a generator
  - inputs, targets, and sample\_weights list which can be used to evaluate the loss and metrics for any model after any epoch has ended.
- validation\_steps: only if the validation\_data is a generator then only this
  argument can be used. It specifies the total number of steps taken from
  the generator before it is stopped at every epoch and its value is
  calculated as the total number of validation data points in your dataset
  divided by the validation batch size.

## TRAIN THE MODEL



## Save The Model

The model is saved with .h5 extension as follows

An H5 file is a data file saved in the Hierarchical Data Format (HDF). It contains multidimensional arrays of scientific data.

## **SAVE THE MODEL**



## Test The Model

- Evaluation is a process during the development of the model to check whether the model is the best fit for the given problem and corresponding data.
- Load the saved model using load\_model
- Taking an image as input and checking the results
- By using the model we are predicting the output for the given input image
- The predicted class index name will be printed here.

## **TEST THE MODEL**