

YILDIZ TECHNICAL UNIVERSITY ELECTRICAL- ELECTRONICS FACULTY COMPUTER ENGINEERING DEPARTMENT

Image Processing Lecture Semester Project

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The content of this report is creating CNN arhitecture and Realization of Transfer Learning.

CNN MODELS

In this project, i have used colab for model training. In order to get data from kaggle i need to get kaggle API and run these cells.

Download data set from kaggle with API

```
In order to download data do these processes:
    1.create API token in kaggle
    2.upload it to the workspace
    3.run all cells

In [4]: %cp /content/kaggle.json /root/.kaggle/

In [3]: %mkdir /root/.kaggle/

In [5]: !kaggle datasets download -d mengcius/cinic10

Warning: Your Kaggle API key is readable by other users on this system! To fix this, you can run 'chmod 600 /root/.kaggle/kaggle.json'
    Downloading cinic10.zip to /content
    99% 745M/754M [00:17<00:00, 44.3MB/s]
    100% 754M/754M [00:17<00:00, 44.7MB/s]

In [6]: import shutil
    shutil.unpack_archive("cinic10.zip", "/content")</pre>
```

In the design of arhitecture of CNN i have used keras library. After importing libraries , parameter initialization starts

Design of CNN arhitecture

input_shape = (IMAGE_WIDTH, IMAGE_HEIGHT, 3)

Read Data Set

```
In [1]: #Import libraries
        %matplotlib inline
        import matplotlib.pyplot as plt
        import matplotlib.image as mpimg
        from keras.preprocessing.image import ImageDataGenerator
        from keras.models import Sequential, Model
        from keras.optimizers import RMSprop
        from keras.layers import Activation, Dropout, Flatten, Dense, GlobalMaxPooling2D
        from keras.callbacks import CSVLogger
        import numpy as np
        import pandas as pd
        import cv2
        import os
In [2]: # Initialize hyperparameters
        IMAGE_SIZE = 224
        IMAGE_WIDTH, IMAGE_HEIGHT = IMAGE_SIZE, IMAGE_SIZE
        TEST_SIZE = 30
```

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CNN - Jupyter Notebook

```
In [3]: from tqdm import tqdm
        #get training data folder location
        traindir = "/content/train"
        #get test data folder location
        testdir = "/content/test"
        #create categories which values are classes of images
        categories = ["airplane", "automobile", "bird", "cat", "deer", "dog", "frog", "horse",
        #create training data list in order to store images
        training data = []
        iterations = 0
        test data = []
        #loop over image folder and get 500 images from each folder
        #because during the training process ram crash error occurs
        #in order to solve this problem i have used 500 images from each class
        for category in categories:
            #get training path
            path = os.path.join(traindir, category)
            class_num = categories.index(category)
            for img in tqdm(os.listdir(path)):
              iterations += 1
              if iterations > 500:
              img_array = cv2.imread(os.path.join(path,img))
              new_array = cv2.resize(img_array, (IMAGE_SIZE, IMAGE_SIZE))
              training_data.append([new_array, class_num])
            iterations = 0
        #create test data with 500 images from each class
        for category in categories:
            path = os.path.join(testdir, category)
            class_num = categories.index(category)
            for img in tqdm(os.listdir(path)):
              iterations += 1
              if iterations > 500:
              img_array = cv2.imread(os.path.join(path,img))
              new_array = cv2.resize(img_array, (IMAGE_SIZE, IMAGE_SIZE))
              test_data.append([new_array, class_num])
            iterations = 0
        print()
        print(len(training_data))
        print(len(test_data))
```

Shuffle training data in order to get proper results in model

```
In [4]: #shuffle training data
import random
random.shuffle(training_data)

for sample in training_data[:10]:
    print(sample[1])
```

Creaate training and test variables in order to fit model

Arhitecture of models in the table below:

2 Katman	Her katman için 32 filtre	3x3 filtre boyutlu	GlorotNormal (xavier_normal_)	ReLu Aktivasyon	0.2 Dropout Oranı	Adam Optimizasyon Algoritması
2 Katman	Her katman için 32 filtre	5x5 filtre boyutlu	GlorotNormal (xavier_normal_)	ReLu Aktivasyon	0.2 Dropout Oranı	Adam Optimizasyon Algoritması
2 Katman	Her katman için 32 filtre	3x3 filtre boyutlu	GlorotNormal (xavier_normal_)	ReLu Aktivasyon	0.7 Dropout Oranı	Adam Optimizasyon Algoritması
2 Katman	Her katman için 32 filtre	3x3 filtre boyutlu	GlorotNormal (xavier_normal_)	ReLu Aktivasyon	0.2 Dropout Oranı	Adam Optimizasyon Algoritması
2 Katman	Her katman için 32 filtre	5x5 filtre boyutlu	GlorotNormal (xavier_normal_)	ReLu Aktivasyon	0.2 Dropout Oranı	Adam Optimizasyon Algoritması
2 Katman	Her katman için 32 filtre	3x3 filtre boyutlu	GlorotNormal (xavier_normal_)	ReLu Aktivasyon	0.7 Dropout Oranı	Adam Optimizasyon Algoritması

```
In [11]: #Create CNN model with Keras
         MODEL 1 includes :
             # 2 layers
             # 32 filter for each layer
             # 3x3 filter size
             # 0.2 Dropout rate
         model = Sequential()
         model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', input_shape=input_shape, activation='relu'))
         model.add(Dropout(0.2))
         model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', activation='relu'))
         model.add(Dropout(0.2))
         model.add(Flatten())
         model.add(Dense(len(categories), activation="softmax"))
         model.compile(loss='categorical_crossentropy', optimizer="adam", metrics=['accuracy'])
         model.summary()
         model.fit(X_train, y_train, batch_size=32, epochs=10)
In [12]: #Create CNN model with Keras
```

```
In [12]: #Create CNN model with Keras
...

MODEL 2 includes :
    # 2 layers
    # 32 filter for each layer
    # 5x5 filter size
    # 0.2 Dropout rate
...

model = Sequential()

model.add(Conv2D(32, (5,5), kernel_initializer='GlorotNormal', input_shape=input_shape, activation='relu'))

model.add(Dropout(0.2))

model.add(Conv2D(32, (5,5), kernel_initializer='GlorotNormal', activation='relu'))

model.add(Dropout(0.2))

model.add(Flatten())
model.add(Dense(len(categories), activation="softmax"))
In [14]: #Create CNN model with Keras
...

#CREATE TABLE TABLE
```

```
In [11]: #Create CNN model with Keras
           MODEL 4 includes :
               # 2 layers
               # 32 filter for each layer
               # 3x3 filter size
               # 0.2 Dropout rate
           model = Sequential()
           model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', input_shape=input_shape, activation='relu'))
           model.add(Dropout(0.2))
           model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', activation='relu'))
           model.add(Dropout(0.2))
           model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', activation='relu'))
           model.add(Dropout(0.2))
           model.add(Flatten())
           model.add(Dense(len(categories), activation="softmax"))
           model.compile(loss='categorical_crossentropy', optimizer="adam", metrics=['accuracy'])
           model.summary()
           model.fit(X_train, y_train, batch_size=32, epochs=10)
 In [12]: #Create CNN model with Keras
          MODEL 5 includes :
              # 2 layers
              # 32 filter for each layer
              # 5x5 filter size
              # 0.2 Dropout rate
          model = Sequential()
          model.add(Conv2D(32, (5,5), kernel initializer='GlorotNormal', input shape=input shape, activation='relu'))
          model.add(Dropout(0.2))
          model.add(Conv2D(32, (5,5), kernel initializer='GlorotNormal', activation='relu'))
          model.add(Dropout(0.2))
          model.add(Conv2D(32, (5,5), kernel_initializer='GlorotNormal', activation='relu'))
          model.add(Dropout(0.2))
          model.add(Flatten())
          model.add(Dense(len(categories), activation="softmax"))
          model.compile(loss='categorical_crossentropy', optimizer="adam", metrics=['accuracy'])
          model.summary()
          model.fit(X_train, y_train, batch_size=32, epochs=10)
In [13]: #Create CNN model with Keras
         MODEL 6 includes :
            # 2 layers
             # 32 filter for each layer
             # 3x3 filter size
             # 0.7 Dropout rate
         model = Sequential()
         model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', input_shape=input_shape, activation='relu'))
         model.add(Dropout(0.7))
         model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', activation='relu'))
         model.add(Dropout(0.7))
         model.add(Conv2D(32, (3,3), kernel_initializer='GlorotNormal', activation='relu'))
         model.add(Dropout(0.7))
         model.add(Flatten())
         model.add(Dense(len(categories), activation="softmax"))
         model.compile(loss='categorical crossentropy', optimizer="adam", metrics=['accuracy'])
         model.summary()
         model.fit(X_train, y_train, batch_size=32, epochs=10)
```

VGG MODEL

Remove last layer, add new layer with 1024 neuron and then compile and fit model.

```
In [ ]: import keras
        from keras.applications.vgg16 import VGG16, preprocess_input
        from keras.models import Sequential, Model
        from keras.layers import Dense
        from keras.optimizers import SGD
        import tensorflow as tf
        from keras import utils
In [ ]: vgg_model = VGG16(weights="imagenet") # train with imagenet
        # remove the output layer
        vgg_model = Model(inputs=vgg_model.inputs, outputs=vgg_model.layers[-2].output)
        vgg model.summary()
In [ ]: #Add fully connected layer with 1024 neuron
        fcc = Dense(1024, activation='relu', name='fcc')(vgg_model.layers[-2].output)
        pred = Dense(3, activation='softmax', name='prediction')(fcc)
        my_vgg16 = Model(vgg_model.input, pred)
        for i in range(0, 21):
         my_vgg16.layers[i].trainable=False
        my_vgg16.summary()
In [ ]: _y_train = utils.to_categorical(y_train)
        y_test = utils.to_categorical(y_test)
In [ ]: #Compile model
        my vgg16.compile(optimizer='adam',
                      loss='binary_crossentropy',
                      metrics=["accuracy"])
        my_vgg16.fit(X_train, _y_train, batch_size=64, epochs=10)
```

VGG-16 Training 4 Layer

```
In [ ]: import keras
        from keras.applications.vgg16 import VGG16, preprocess_input
        from keras.models import Sequential, Model
        from keras.layers import Dense
        from keras.optimizers import SGD
        import tensorflow as tf
        from keras import utils
In [ ]: vgg_model = VGG16(weights="imagenet") # train with imagenet
        # remove the output layer
        vgg_model = Model(inputs=vgg_model.inputs, outputs=vgg_model.layers[-2].output)
        vgg_model.summary()
In [ ]: #Add fully connected layer with 1024 neuron
        fcc = Dense(1024, activation='relu', name='fcc')(vgg_model.layers[-2].output)
        pred = Dense(len(categories), activation='softmax', name='prediction')(fcc)
        my_vgg16 = Model(vgg_model.input, pred)
        for i in range(0, 19):
         my_vgg16.layers[i].trainable=False
        my_vgg16.summary()
In [ ]: _y_train = utils.to_categorical(y_train)
        _y_test = utils.to_categorical(y_test)
In [ ]: #Compile model
        my_vgg16.compile(optimizer='adam',
                      loss='binary_crossentropy',
                      metrics=["accuracy"])
        my_vgg16.fit(X_train, _y_train, batch_size=64, epochs=10)
```

RESNET MODEL

Resnet-50 Training 1 Layer

```
In [ ]: from tensorflow.keras.preprocessing import image
        from tensorflow.keras.applications.resnet50 import ResNet50, preprocess input, decode predictions
        from keras.models import Sequential, Model
        from keras.layers import Dense
        from keras import utils
In [ ]: resnet_model = ResNet50(weights='imagenet') #train with imagenet
        resnet model.summary()
In [ ]: # add 1 fully-connected and 1 prediction
        fcc = Dense(1024, name='fcc')(resnet_model.layers[-2].output)
        pred = Dense(len(categories), activation='softmax', name='prediction')(fcc)
        my_resnet50 = Model(resnet_model.input, pred)
        for i in range(0, 175):
         my_resnet50.layers[i].trainable=False
        my_resnet50.summary()
In [ ]: _y_train = utils.to_categorical(y_train)
        _y_test = utils.to_categorical(y_test)
In [ ]: #compile model
        my_resnet50.compile(optimizer='adam',
                      loss='binary crossentropy',
                      metrics=["accuracy"])
        my_resnet50.fit(X_train, _y_train, batch_size=32, epochs=10)
```

Confusion matrix results:

```
In [ ]: #import confusion matrix and get scores
        from sklearn.metrics import classification_report
        from sklearn.metrics import confusion_matrix
        y_pred = my_resnet50.predict(X_test)
        y_pred = np.argmax(y_pred, axis = 1)
        _y_test = np.argmax(_y_test, axis = 1)
        print(confusion_matrix(_y_test,y_pred))
        print(classification_report(_y_test,y_pred))
        [[ 96 214 190]
           6 403 91]
           4 115 381]]
                      precision
                                 recall f1-score
                                                      support
                   0
                           0.91
                                     0.19
                                               0.32
                                                          500
                           0.55
                                     0.81
                                                          500
                   1
                                               0.65
                   2
                           0.58
                                     0.76
                                               0.66
                                                          500
                                               0.59
                                                         1500
            accuracy
                           0.68
                                     0.59
                                                         1500
           macro avg
                                               0.54
        weighted avg
                           0.68
                                     0.59
                                               0.54
                                                         1500
```

Resnet-50 Training 4 Layer

```
In [ ]: from tensorflow.keras.preprocessing import image
        from tensorflow.keras.applications.resnet50 import ResNet50, preprocess_input, decode_predictions
        from keras.models import Sequential, Model
        from keras.layers import Dense
        from keras import utils
In [ ]: resnet_model = ResNet50(weights='imagenet') #training with imagenet
        resnet_model.summary()
        conv2_block2_1_bn (BatchNormali (None, 56, 56, 64)
                                                             256
                                                                         conv2_block2_1_conv[0][0]
        conv2_block2_1_relu (Activation (None, 56, 56, 64)
                                                                         conv2 block2 1 bn[0][0]
        conv2_block2_2_conv (Conv2D)
                                        (None, 56, 56, 64)
                                                             36928
                                                                         conv2_block2_1_relu[0][0]
        conv2_block2_2_bn (BatchNormali (None, 56, 56, 64)
                                                             256
                                                                         conv2_block2_2_conv[0][0]
        conv2_block2_2_relu (Activation (None, 56, 56, 64)
                                                                         conv2_block2_2_bn[0][0]
        conv2_block2_3_conv (Conv2D)
                                        (None, 56, 56, 256)
                                                             16640
                                                                         conv2_block2_2_relu[0][0]
        conv2_block2_3_bn (BatchNormali (None, 56, 56, 256) 1024
                                                                         conv2_block2_3_conv[0][0]
        conv2_block2_add (Add)
                                        (None, 56, 56, 256) 0
                                                                         conv2_block1_out[0][0]
                                                                         conv2_block2_3_bn[0][0]
        conv2_block2_out (Activation)
                                        (None, 56, 56, 256) 0
                                                                         conv2_block2_add[0][0]
In [ ]: # add 1 fully-connected and 1 prediction
        fcc = Dense(1024, name='fcc')(resnet_model.layers[-2].output)
        pred = Dense(len(categories), activation='softmax', name='prediction')(fcc)
        my_resnet50 = Model(resnet_model.input, pred)
        for i in range(0, 173): # only last 4 will be trainable
         my_resnet50.layers[i].trainable=False
        my_resnet50.summary()
```

Confusion matrix results:

```
In [ ]: #import confusion matrix and get scores
        from sklearn.metrics import classification_report
        from sklearn.metrics import confusion_matrix
        y_pred = my_resnet50.predict(X_test)
        y_pred = np.argmax(y_pred, axis = 1)
        _y_test = np.argmax(_y_test, axis = 1)
        print(confusion_matrix(_y_test,y_pred))
        print(classification_report(_y_test,y_pred))
        [[288 188 24]
         [ 77 418 5]
         [208 162 130]]
                      precision recall f1-score
                                                    support
                   0
                          0.50
                                    0.58
                                              0.54
                                                         500
                          0.54
                                    0.84
                                                         500
                   1
                                              0.66
                   2
                          0.82
                                              0.39
                                                         500
                                    0.26
                                              0.56
                                                        1500
            accuracy
                          0.62
                                    0.56
                                              0.53
                                                        1500
           macro avg
        weighted avg
                          0.62
                                    0.56
                                              0.53
                                                        1500
```

Model Weight Load

VGG H5 Model Weight Load

ResNet50 Model Weight Load ¶