

# Create Deep Learning Network Architecture with Pretrained Parameters

Script for creating the layers for a deep learning network with the following properties:

```
Number of layers: 43
Number of connections: 42
Pretrained parameters file: /MATLAB Drive/params_2020_11_22__22_06_14.mat
```

Run the script to create the layers in the workspace variable `layers`.

To learn more, see [Generate MATLAB Code From Deep Network Designer](#).

Auto-generated by MATLAB on 22-Nov-2020 22:06:22

## Load the Pretrained Parameters

```
params = load("/home/basia/Documents/Filtry/params_2020_11_22__22_06_14.mat");
```

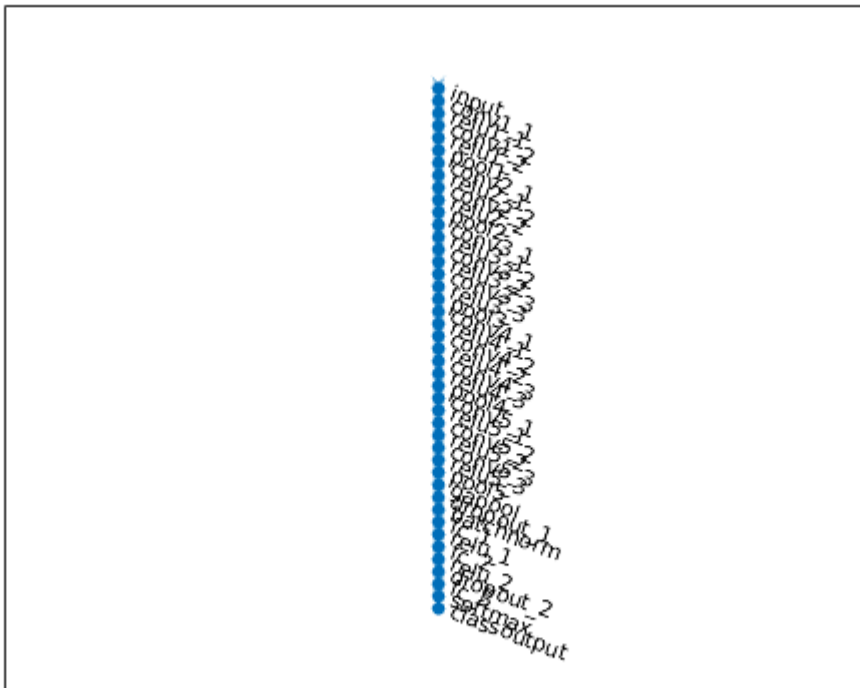
## Create Array of Layers

```
layers = [
    imageInputLayer([224 224 3],"Name","input","Mean",params.input.Mean)
    convolution2dLayer([3 3],64,"Name","conv1_1","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu1_1")
    convolution2dLayer([3 3],64,"Name","conv1_2","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu1_2")
    maxPooling2dLayer([2 2],"Name","pool1","Stride",[2 2])
    convolution2dLayer([3 3],128,"Name","conv2_1","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu2_1")
    convolution2dLayer([3 3],128,"Name","conv2_2","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu2_2")
    maxPooling2dLayer([2 2],"Name","pool2","Stride",[2 2])
    convolution2dLayer([3 3],256,"Name","conv3_1","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu3_1")
    convolution2dLayer([3 3],256,"Name","conv3_2","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu3_2")
    convolution2dLayer([3 3],256,"Name","conv3_3","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu3_3")
    maxPooling2dLayer([2 2],"Name","pool3","Stride",[2 2])
    convolution2dLayer([3 3],512,"Name","conv4_1","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu4_1")
    convolution2dLayer([3 3],512,"Name","conv4_2","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu4_2")
    convolution2dLayer([3 3],512,"Name","conv4_3","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu4_3")
    maxPooling2dLayer([2 2],"Name","pool4","Stride",[2 2])
    convolution2dLayer([3 3],512,"Name","conv5_1","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu5_1")
    convolution2dLayer([3 3],512,"Name","conv5_2","Padding",[1 1 1 1],"WeightL2Factor",0)
    reluLayer("Name","relu5_2")
    convolution2dLayer([3 3],512,"Name","conv5_3","Padding",[1 1 1 1],"WeightL2Factor",0)
```

```
reluLayer("Name", "relu5_3")
maxPooling2dLayer([2 2], "Name", "pool5", "Stride", [2 2])
globalAveragePooling2dLayer("Name", "gapool")
dropoutLayer(0.2, "Name", "dropout_1")
batchNormalizationLayer("Name", "batchnorm")
fullyConnectedLayer(4096, "Name", "fc_1")
reluLayer("Name", "relu_1")
fullyConnectedLayer(4096, "Name", "fc_2")
reluLayer("Name", "relu_2")
dropoutLayer(0.2, "Name", "dropout_2")
fullyConnectedLayer(6, "Name", "fc_3")
softmaxLayer("Name", "softmax")
classificationLayer("Name", "classoutput"]];
```

## Plot Layer

```
plot(layerGraph(layers));
```



## Import Data

```
data = imageDatastore('/home/basia/Documents/Filtry/titan/', ...
    'IncludeSubfolders', true, ...
    'LabelSource', 'foldernames');
imageSize = layers(1).InputSize;
%data.ReadFcn = @(data)imresize(imread(data),imageSize);
[trainingSet, testSet] = splitEachLabel(data, 0.2);
```

```

auimdsTrain = augmentedImageDatastore(imageSize,trainingSet,'ColorPreprocessing','gray2rgb')
auimdsTest = augmentedImageDatastore(imageSize,testSet,'ColorPreprocessing','gray2rgb')

```

```

auimdsTrain.preview()

```

```

ans = 8x2 table

```

	input	response
1	224x224x3 ... AML	
2	224x224x3 ... AML	
3	224x224x3 ... AML	
4	224x224x3 ... AML	
5	224x224x3 ... AML	
6	224x224x3 ... AML	
7	224x224x3 ... AML	
8	224x224x3 ... AML	

## Training Options

```

options = trainingOptions('rmsprop', "MaxEpochs",10, 'MiniBatchSize', 32, "Plots","trainAndValidatePlots');

```

## Train

```

tumorNet = trainNetwork(auimdsTrain, layers, options);

```

Training on single GPU.

Initializing input data normalization.

Epoch	Iteration	Time Elapsed (hh:mm:ss)	Mini-batch Accuracy	Validation Accuracy	Mini-batch Loss	Validation Loss	Batch Size
1	1	00:00:14	6.25%	51.93%	1.9409	7.2600	32
1	10	00:00:52	43.75%	53.44%	1.4737	1.4507	32
2	20	00:01:33	46.88%	51.63%	1.5931	1.3076	32
3	30	00:02:14	43.75%	51.85%	1.3263	1.3112	32
4	40	00:02:55	46.88%	51.63%	1.2692	1.4397	32
5	50	00:03:36	53.12%	51.55%	1.2751	2.3018	32
6	60	00:04:19	53.12%	51.93%	1.3269	1.3951	32
7	70	00:05:00	50.00%	51.93%	1.2849	1.3923	32
8	80	00:05:39	43.75%	51.93%	1.3671	1.4694	32
9	90	00:06:09	50.00%	51.93%	1.2438	1.2982	32
10	100	00:06:50	50.00%	52.61%	1.2802	1.3621	32