MCDB 4110/6440 – Quantitative Microscopy Lab

Lab 4 – 2:

Evaluating a network and semantic segmentation

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Recap of last week

Activation function



A typical training cycle



SGDM computes the derivative of the error, then changes the weights based on this slope



Neural networks have connected layers of perceptrons



Pooling is used to combine pixels in the feature maps (e.g., combine 2x2 pixels into 1)



Convolutional neural networks have several layers of convolutional layers



Each convolutional layer combines features from the previous







Layer 1 Detect lines and edges Layer 2 Combine lines and edges to detect eyes, ears, noses

Layer 3 Combine eyes, ears, noses to detect faces



Results

Validation accuracy:	83.43%
Training finished:	Reached final iteration
Training Time	
Start time:	09-Nov-2021 19:15:42
Elapsed time:	55 sec
Training Cycle	
Epoch:	5 of 5
Iteration:	215 of 215
Iterations per epoch:	43
Maximum iterations:	215
Validation	
Frequency:	50 iterations
Other Information	
Hardware resource:	Single GPU
Learning rate schedule:	Constant
Learning rate:	0.001
i Learn more	

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pizza











hamburger









pizza









hamburger



french_fries



hamburger







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Deep Learning Visualization

Activations

 To understand what filters have been trained, you can use the function activations to extract learned image features from a trained convolutional network

Code

```
Irgb = imread('D:\Teaching\IQBioLabs_2021\Week
1\bpae_fullcell.tif');
```

```
act1 = activations(alexnet_lab4, Irgb, 'conv1');
sz = size(act1);
act1 = reshape(act1,[sz(1) sz(2) 1 sz(3)]);
I = imtile(mat2gray(act1),'GridSize',[5 5]);
```

imshow(I)

Example: First 25 activations from Layer 1





The activations show features that were detected



Plot the strongest activation

[maxValue,maxValueIndex] = max(max(max(act1)));

act1chMax = act1(:,:,:,maxValueIndex);

act1chMax = mat2gray(act1chMax);

act1chMax = imresize(act1chMax,[256 256]);

I = imtile({Irgb,act1chMax});

Plot the strongest activation



Plot the strongest activation



Confusion Matrix

 A confusion matrix shows the predicted vs expected classifications from the trained network

							-				
	airplane	923	4	21	8	4	1	5	5	23	6
ß	automobile	5	972	2					1	5	15
	bird	26	2	892	30	13	8	17	5	4	3
n	cat	12	4	32	826	24	48	30	12	5	7
(CDIC)	deer	5	1	28	24	898	13	14	14	2	1
2	dog	7	2	28	111	18	801	13	17		3
	frog	5		16	27	3	4	943	1	1	
	horse	9	1	14	13	22	17	3	915	2	4
	ship	37	10	4	4		1	2	1	931	10
	truck	20	39	3	3			2	1	9	923
airplane bird cat deer dog trog horse ship tru											unck
	Basel start Olars										

The blue cells are the correct labels

Predicted Class

How to visualize the confusion matrix

- Read in the validation images into an Image Datastore
- Use the classify method to generate the predicted labels
- Use the function confusionchart to display the matrix

confusionchart(trueLabels, predictedLabels)

Try this

 Let's try this using either your own network or using the one I supplied



Semantic segmentation

Semantic segmentation using U-net



https://arxiv.org/pdf/1505.04597.pdf

Semantic segmentation using U-net



Setting up the Unet in MATLAB

lgraph = unetLayers(imageSize, numClasses)

Image augmentation

- To successfully train the U-net, you need 10k 100k images
- But we clearly do not have that
- One way to get more images to use "image augmentation":
 - Rotate, translate, reflect images that we have









Test semantic segmentation

- I = readimage(imds, 10);
- C = semanticseg(I, net);

B = labeloverlay(I, C); imshow(B)

Evaluating the results

pxdsResults = semanticseg(imds, net); metrics = evaluateSemanticSegmentation(... pxdsResults, pxdsTruth)