

Mondays with MATLAB

Hands on workshop

Deep Learning on Alvis - the AI Cluster



Emelie Andersson, *Application Engineer*, **MathWorks**

Sagar Zade, *Customer Success Engineer*, **MathWorks**

Rohit Agrawal, *Customer Success Engineer*, **MathWorks**

Anders Sjöström, *National Co-ordinator and Application Expert at SNIC*, **Lund University**

How can you participate?

You have three options:

1. Run exercises yourself on Alvis (you need to have SNIC and Alvis credentials)
2. Run exercises on your own computer (if you don't have access to Alvis, but access to MATLAB)

Download exercise material: <https://tinyurl.com/4je3v6vm>

Username: AppleHill

Password: fjr5n8nf

3. Get your popcorn and just follow along



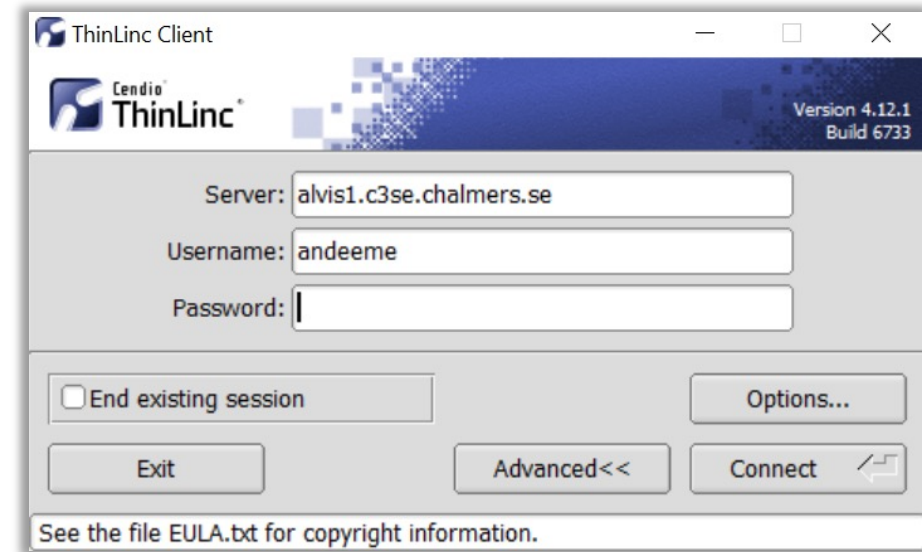
What's Alvis?

“The Alvis cluster is a national SNIC resource dedicated for **Artificial Intelligence** and **Machine Learning** research. The system is built around Graphical Processing Units (GPUs) accelerator cards and consists of several types of compute nodes with multiple Nvidia GPUs.”



Set-up Instructions for Alvis

- Login to Chalmers VPN
- Login to Alvis using Thinlinc
 - alvis1.c3se.chalmers.se
 - Alvis specific username and password



- Open terminal window
 - Copy workshop material:
 - `cp -r /cephyr/users/andeeme/ExerciseFilesDLWorkshop .`
 - **Get on a compute node:**
 - `interactive --reservation=matlab-monday -n8 -t 03:00:00`
- Start MATLAB (from terminal)
 - `module load MATLAB/R2021a`
 - `srun -A SNIC2021-7-61 -t 03:00:00 -gpus-per-node=T4:1 --x11 -pty matlab`

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Introduction



Exercise 1: Deep learning in 6 lines of code

Deep Learning Fundamentals

Break



Exercise 2 & 3: Classify blood smear images

Break



Exercise 4: Improving Network Accuracy

Conclusion

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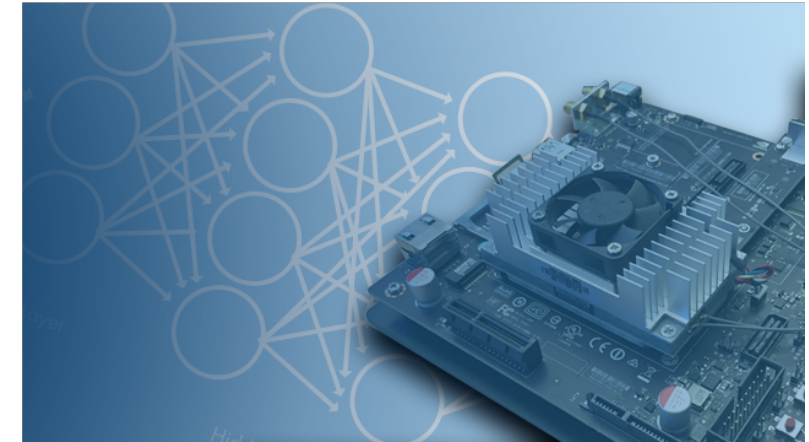
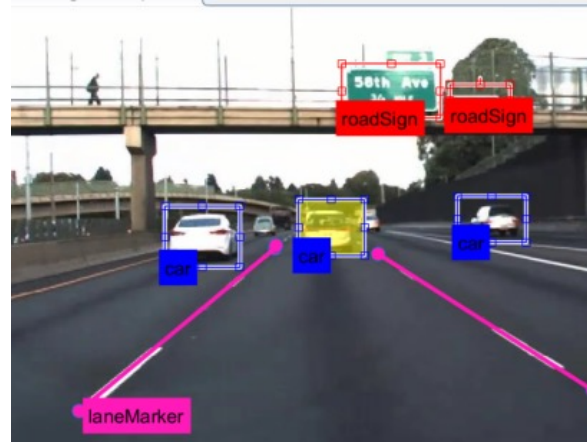
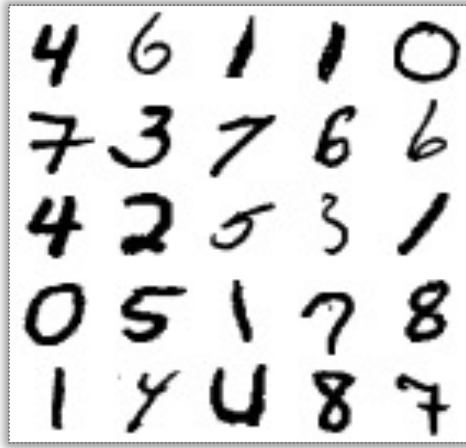
Exercise 2 & 3: Classify blood smear images

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Conclusion



Deep Learning?



12 40.0%	0 0.0%	100% 0.0%
0 0.0%	18 60.0%	100% 0.0%
100% 0.0%	100% 0.0%	100% 0.0%

Some Basic Definitions

Artificial Intelligence

Machine Learning

Deep Learning

Some Basic Definitions

Artificial Intelligence

The ability of

Machine Learning

The practice of **learning a task from data**
without relying on a predetermined equation or model

Some Basic Definitions

Artificial Intelligence

Machine Learning

Deep Learning

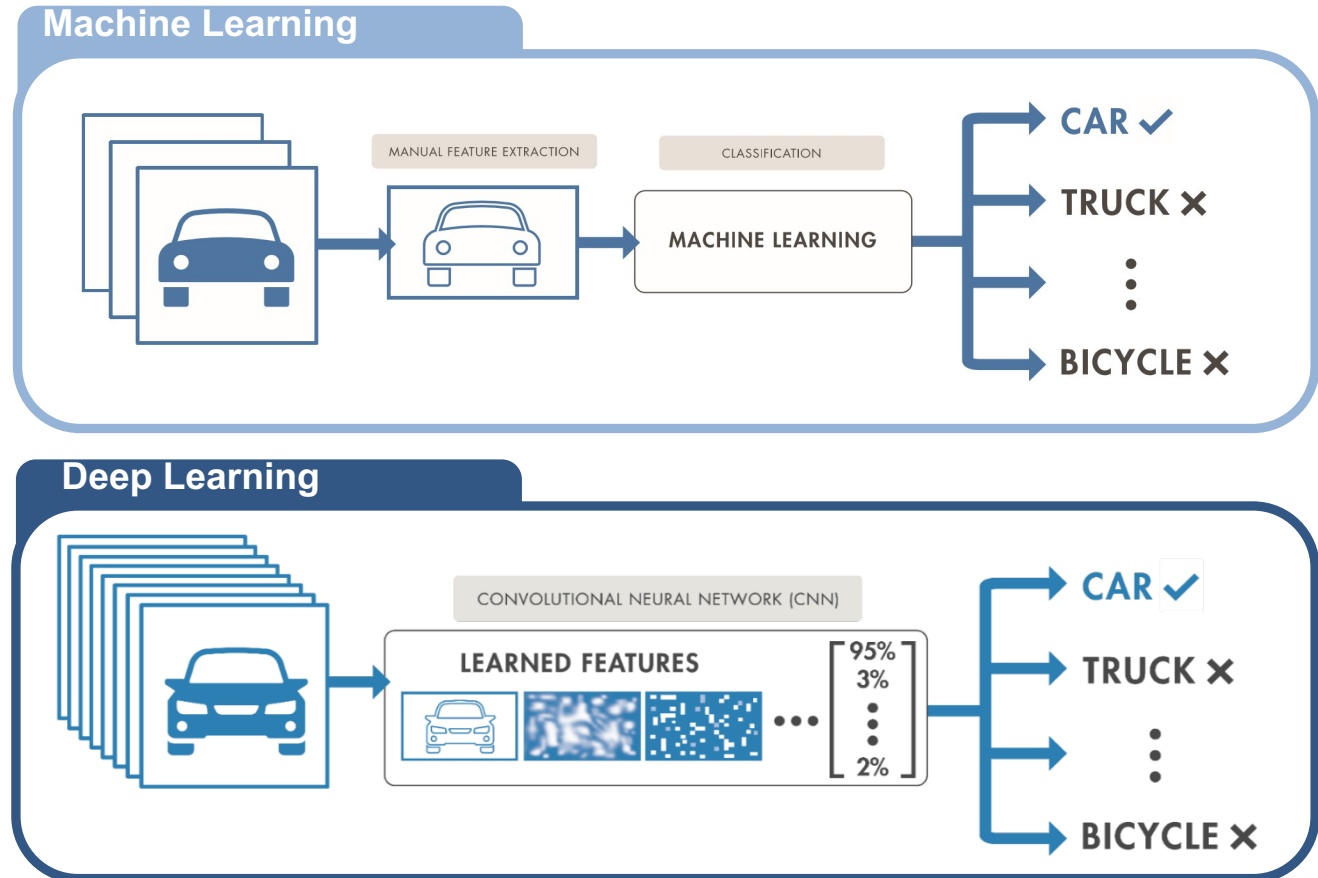
A **type** of machine learning based on
neural networks

Machine Learning & Deep Learning

- Ability to learn from data inherently without being explicitly programmed
 - Learns complex non-linear relationship
 - More Data = better model

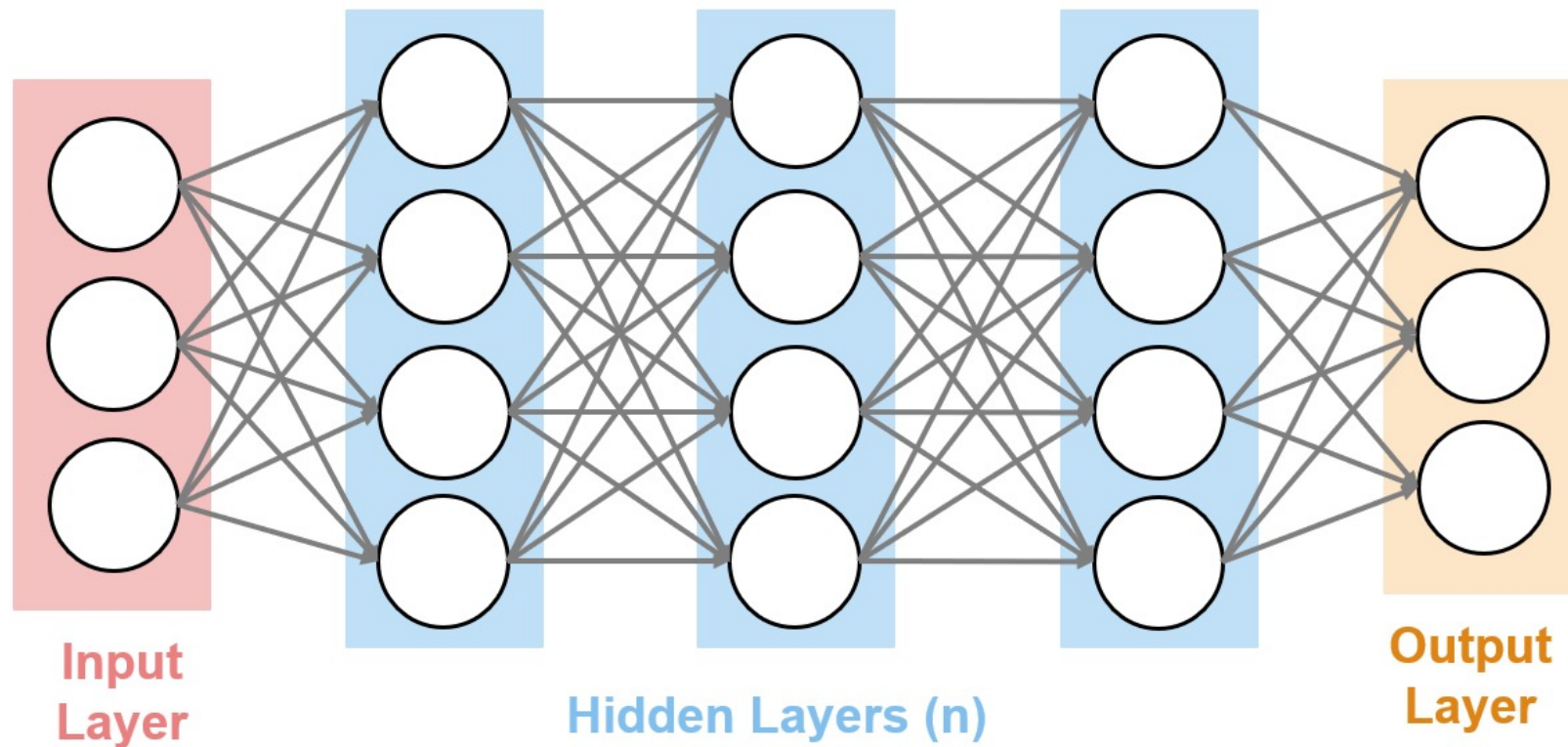
**Machine
Learning**

**Deep
Learning**



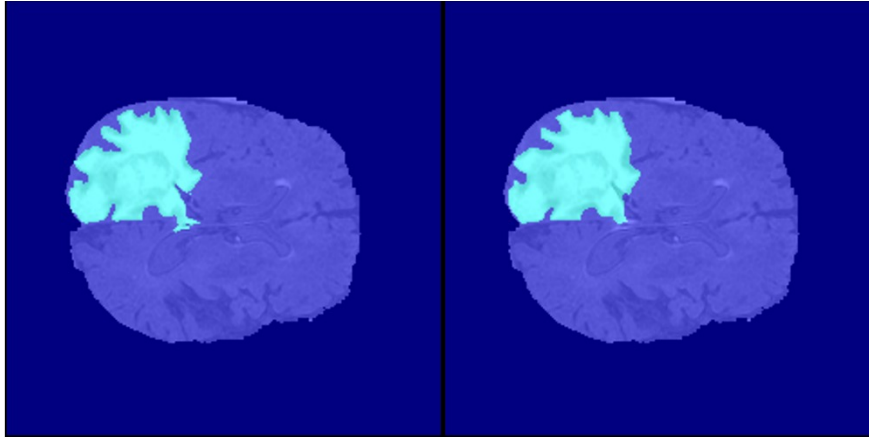
Neural Network Structure

- Deep neural networks have many layers
- Data is passed through the network, and the layer parameters are updated (training)

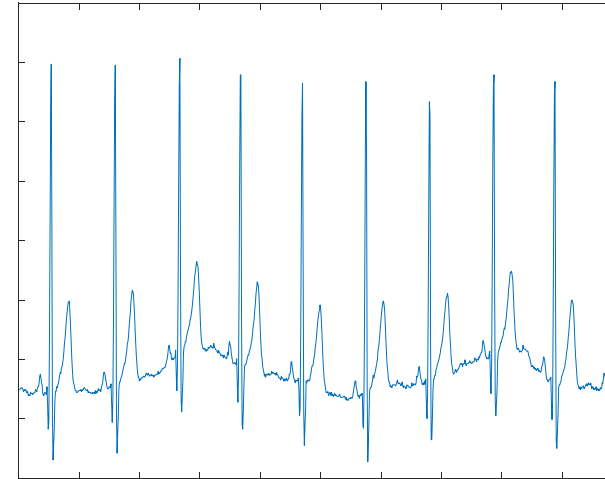


Machine Learning and Deep Learning Datatypes

Image



Signal



Numeric

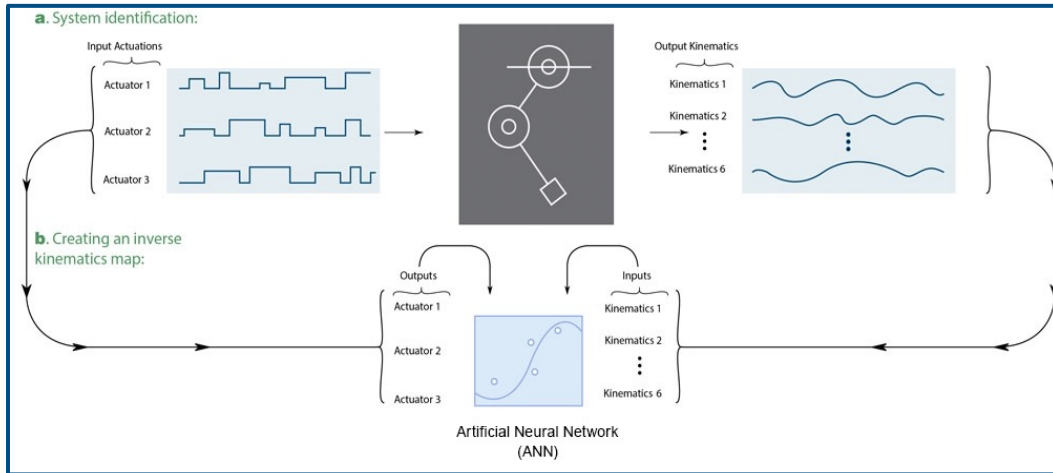
AgeCat	WeightQ	GroupCount	mean_BloodPressure	
Under 30	Q1	6	123.17	79.667
Under 30	Q2	3	120.33	79.667
Under 30	Q3	2	127.5	86.5
Under 30	Q4	4	122	78
30-39	Q1	12	121.75	81.75
30-39	Q2	9	119.56	82.556
30-39	Q3	9	121	83.222
30-39	Q4	11	125.55	87.273
Over 40	Q1	7	122.14	84.714
Over 40	Q2	13	123.38	79.385
Over 40	Q3	14	123.07	84.643
Over 40	Q4	10	124.6	85.1

Text

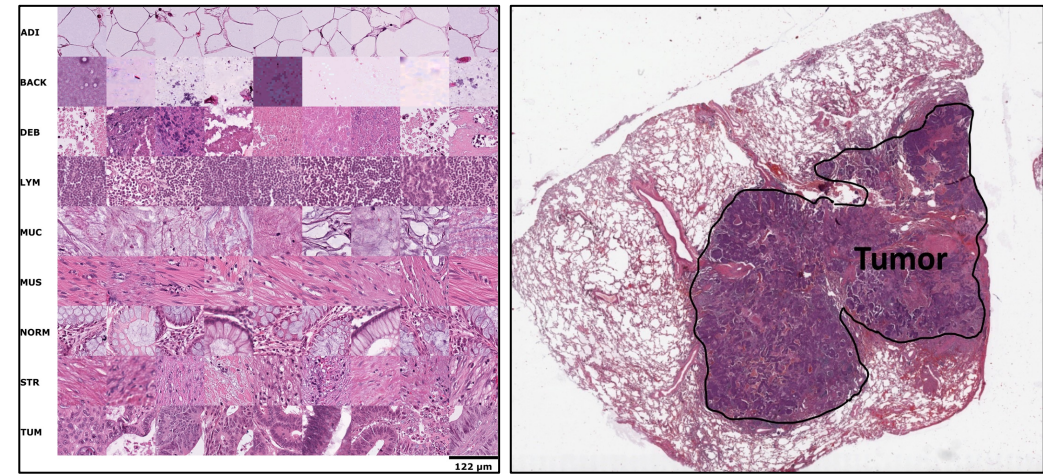


Deep Learning and AI in Research

University of
Southern California



Reinforcement Learning for Robotic Arm



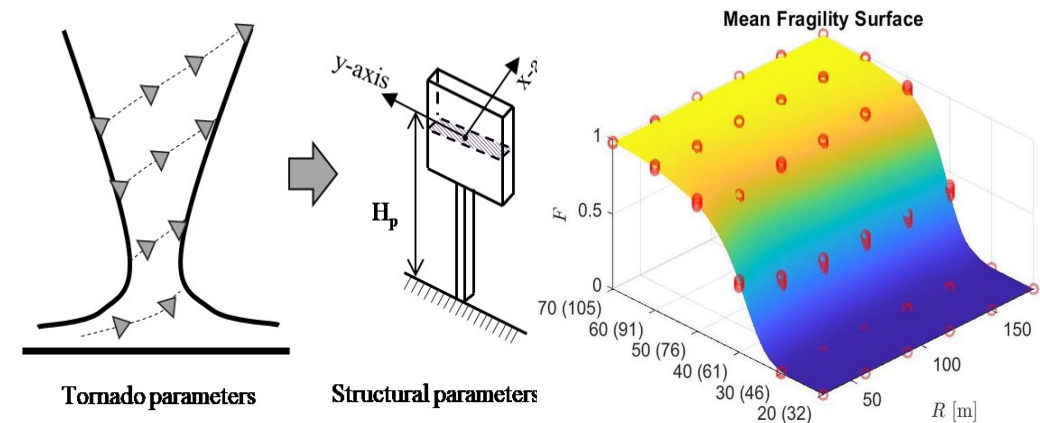
Deep Learning for Tumor Detection

DKFZ
Heidelberg

University of
Twente



Augmented Reality of blood flow



Neural Networks simulate tornadic wind load

Northeast
University

Deep Learning Demo

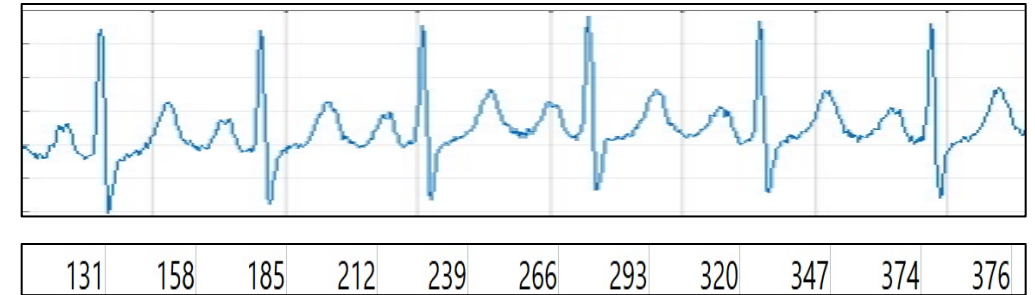
Image Classification

Deep Learning Networks Take in Numeric Data



199	206	208	201	188	178	165	164	180
202	205	202	188	176	169	178	186	183
203	206	189	178	181	183	182	154	87
203	192	184	186	177	167	153	181	192
191	182	176	166	153	141	136	180	227
166	165	154	154	138	137	169	170	211
158	150	145	183	144	156	158	154	179
143	51	98	144	129	130	143	178	123
107	50	33	95	152	173	192	159	87
104	100	84	120	132	172	131	64	94
119	101	97	81	90	109	87	106	111
127	122	110	97	108	120	133	131	134
111	117	108	119	131	143	146	141	156
126	122	113	119	139	142	155	161	151
129	126	130	111	103	130	149	149	156
138	128	136	144	136	129	134	122	145
154	133	134	141	168	150	126	127	151

Images are a numeric matrix



Signals are numeric vectors

The Bird Flies = [0 13 5 6]
 The Leaf Is Brown = [13 3 11 2]

Text is processed as numeric vectors

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Exercise 2 & 3: Classify blood smear images

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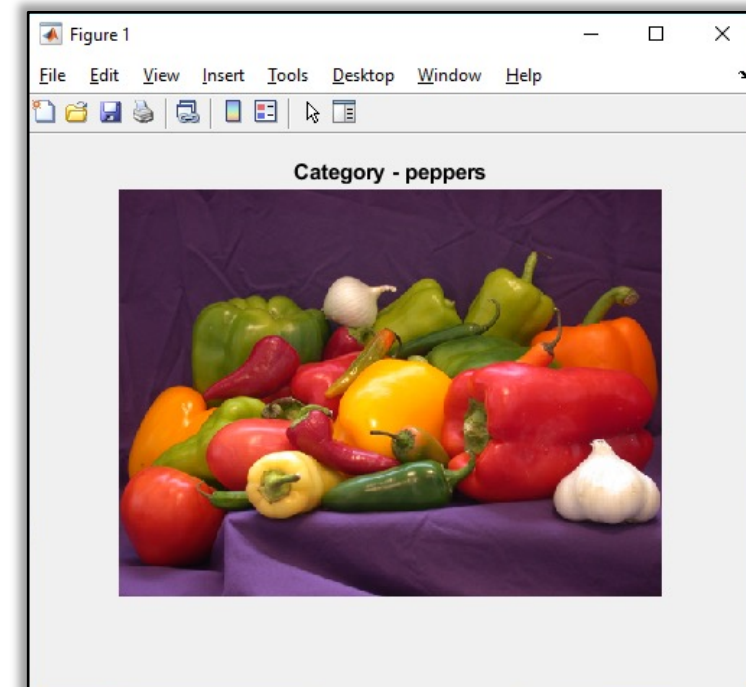
Exercise 4: Improving Network Accuracy

Conclusion

Exercise 1 – Deep Learning in 6 Lines of Code

- Run **runMeFirst.mlx**
- Open **work_deeplearningin6lines.mlx** in **01-DeepLearning101** folder
- Add `load alexnet.mat` at the top to read network
- If you have an account on Alvis, or running on your own computer – Go to the **breakout room** and finish the exercise
- If you don't have a MATLAB in front of you – follow along in Main Room. Let's do the exercise together!

```
% Load the AlexNet neural network  
load alexnet.mat;  
net = alexnet;
```



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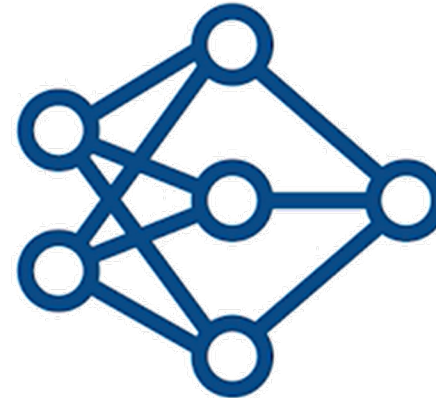


Exercise 4: Improving Network Accuracy

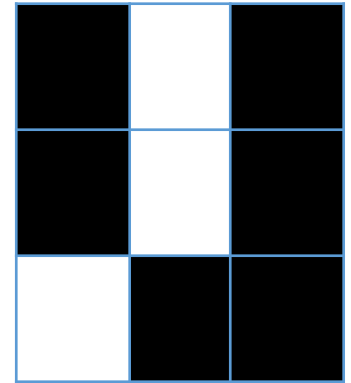
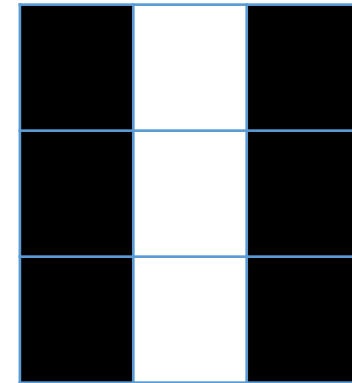
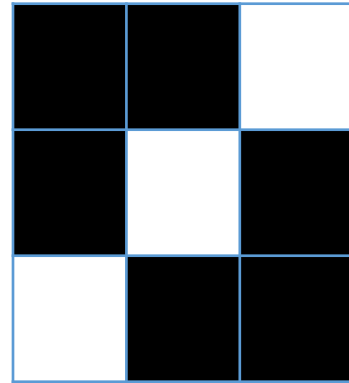
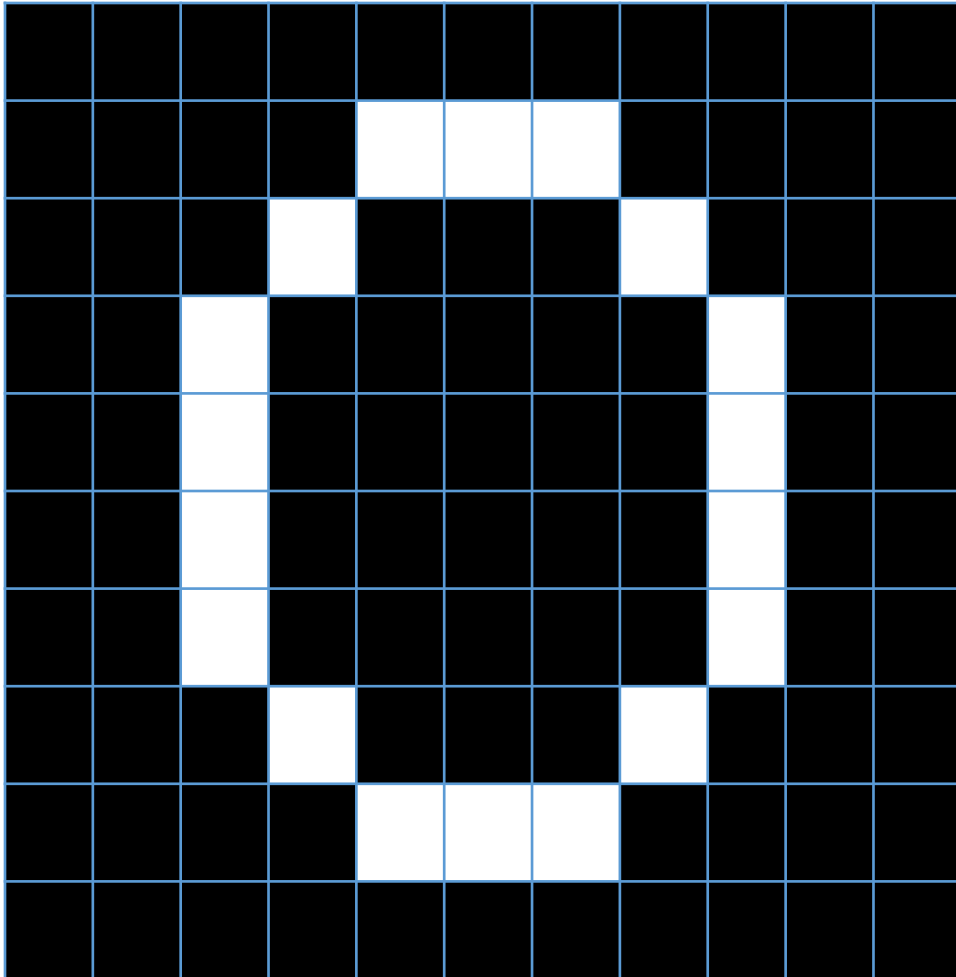
Conclusion

Creating Layer Architectures

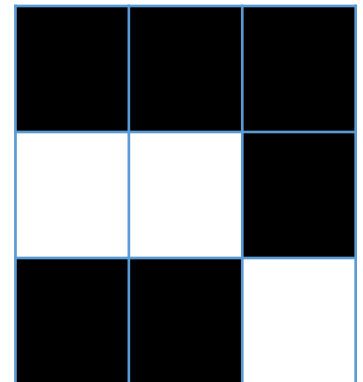
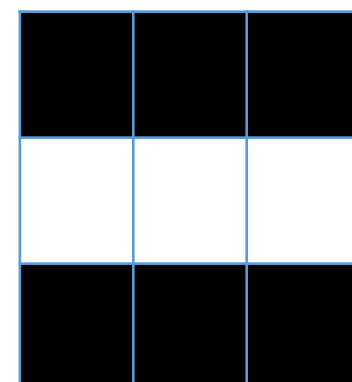
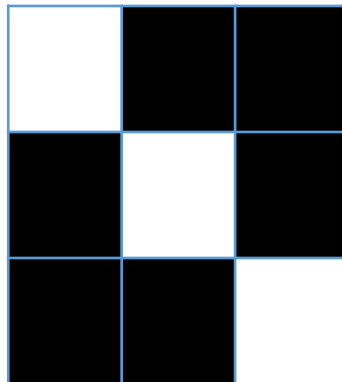
- Convolutional Neural Networks – CNN
- Special layer combinations which learn parameters to classify images
- Convolution Layer
- ReLU Layer
- Max Pooling Layer



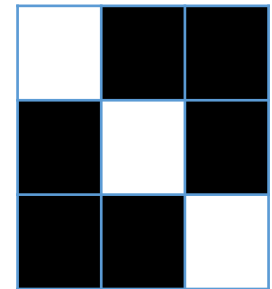
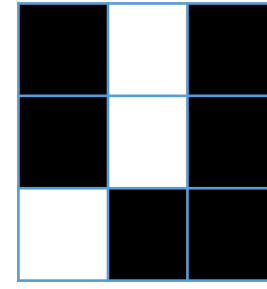
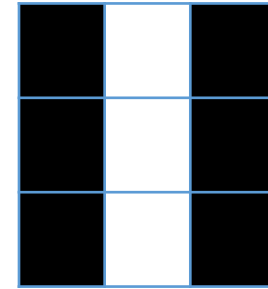
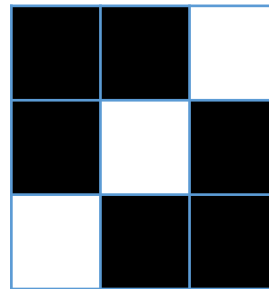
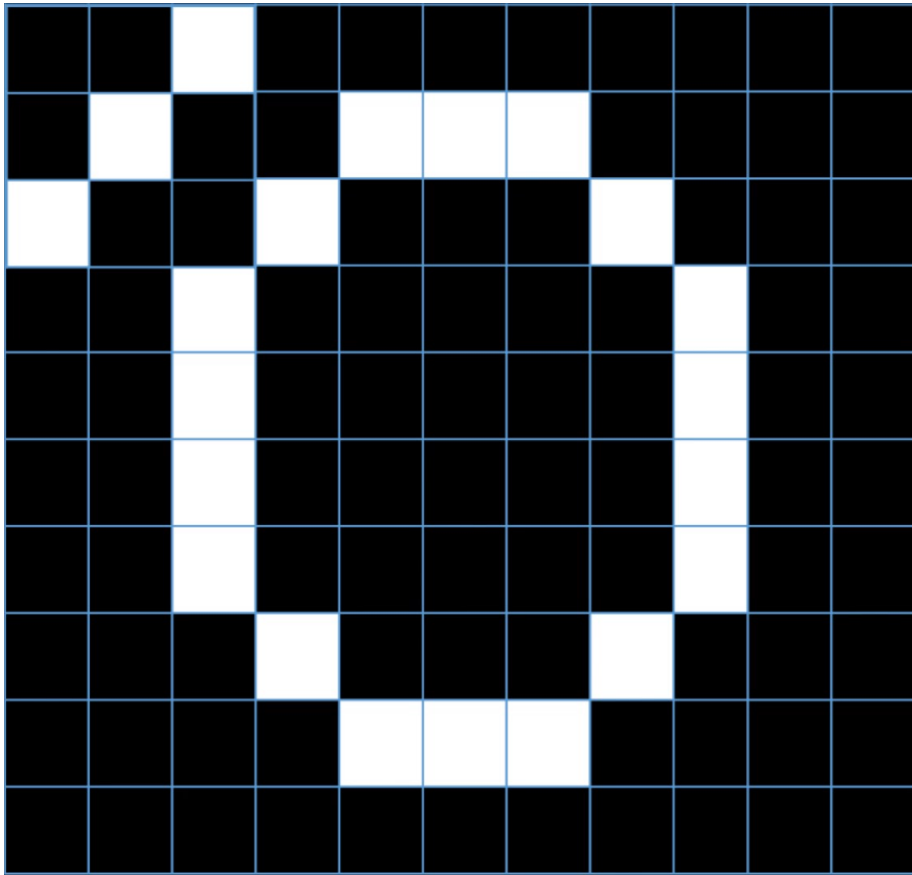
Convolution Layers Search for Patterns



These patterns would be common in the number 0



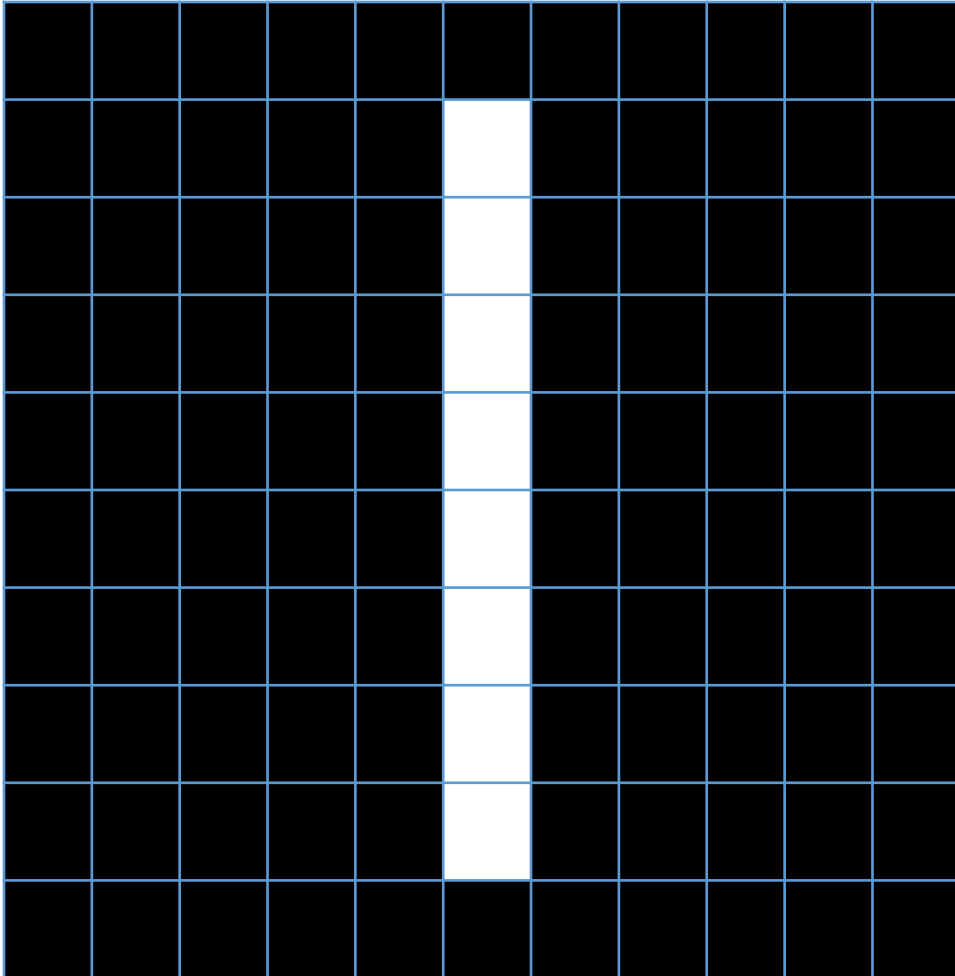
All patterns are compared to the patterns on a new image



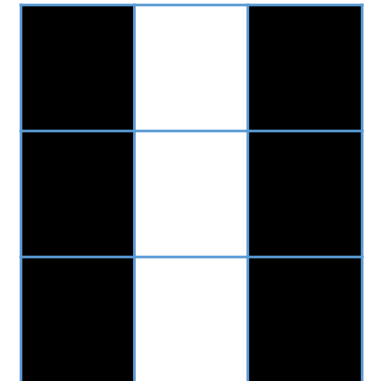
...

- Pattern starts at left corner
Perform comparison
Slide over one pixel
- Reach end of image
- Repeat for next pattern

Good pattern matching in convolution improves chances that object will classify properly



- This image would not match well against the patterns for the number zero
- It would only do very well against this pattern



Activation: Rectified Linear Units Layer (ReLU)

Converts negative numbers to zero

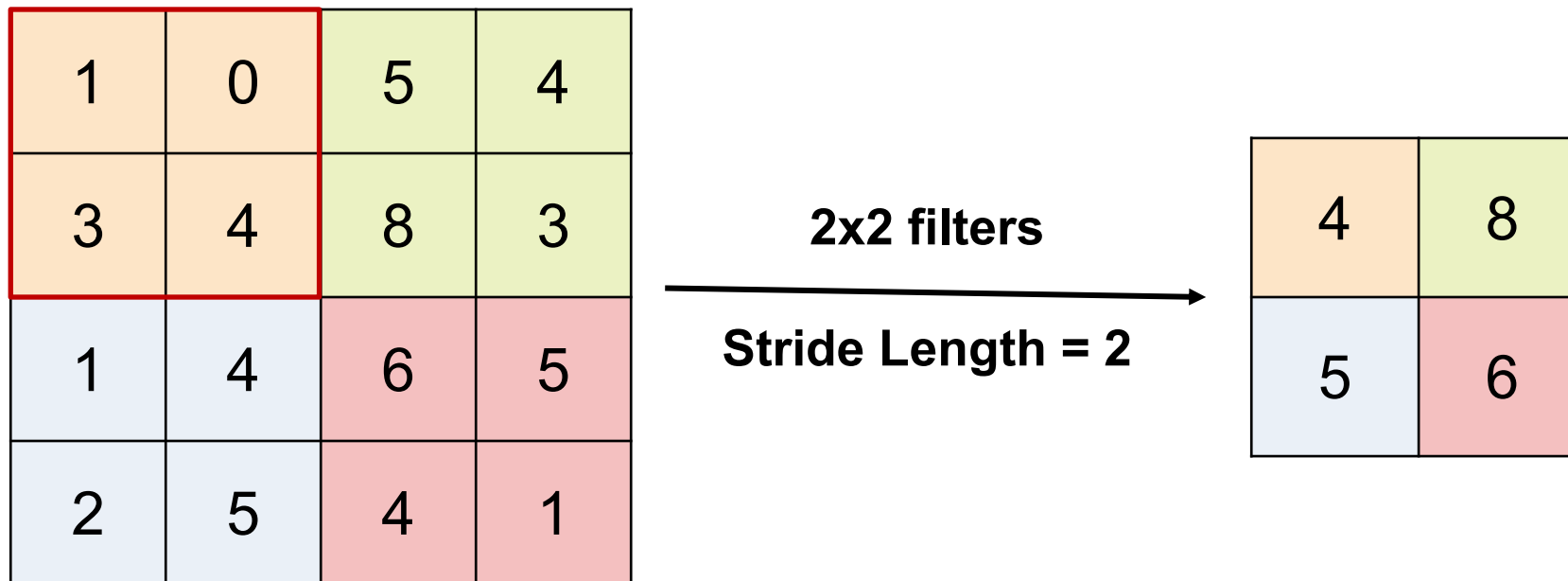
-1	0	5	4
3	-4	-8	3
1	4	6	-5
-2	-5	4	1



0	0	5	4
3	0	0	3
1	4	6	0
0	0	4	1

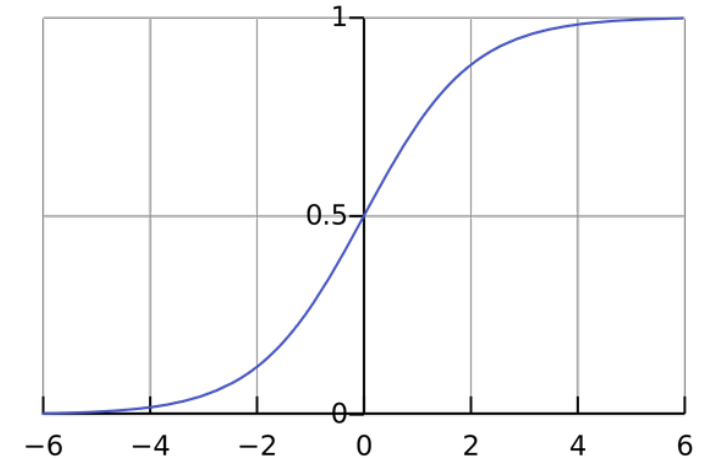
Max Pooling is a down-sampling operation

Shrink large images while preserving important information



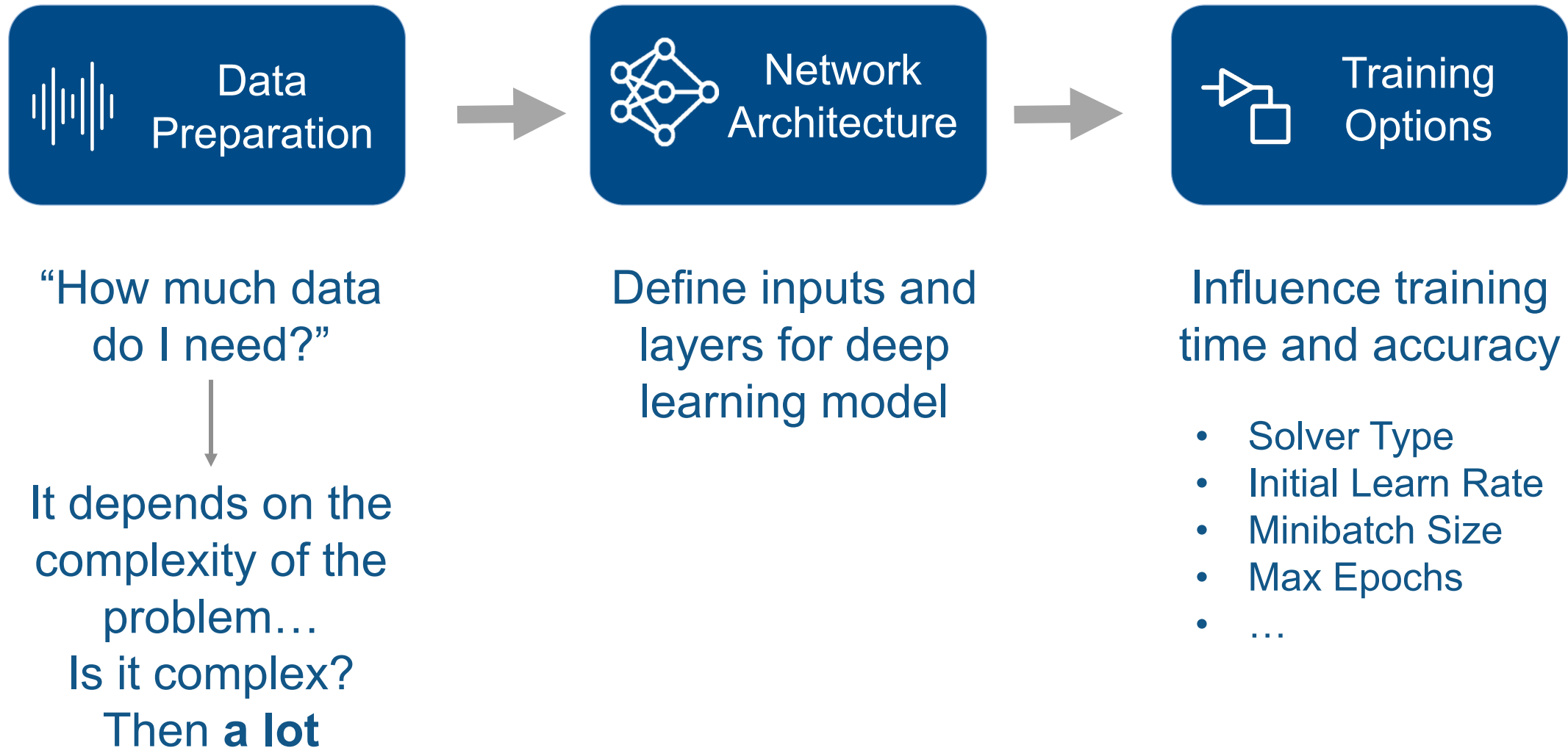
Classification Problems End with 3 Layers

- Fully Connected Layer
 - Looks at which high-level features correspond to a specific category
 - Calculates scores for each category (highest score wins)
- Softmax Layer
 - Turns scores into probabilities.
- Classification Layer
 - Categorizes image into one of the classes that the network is trained on



Note: Regression problems end with a fully connected layer and regression layer

3 Components to Train any Network



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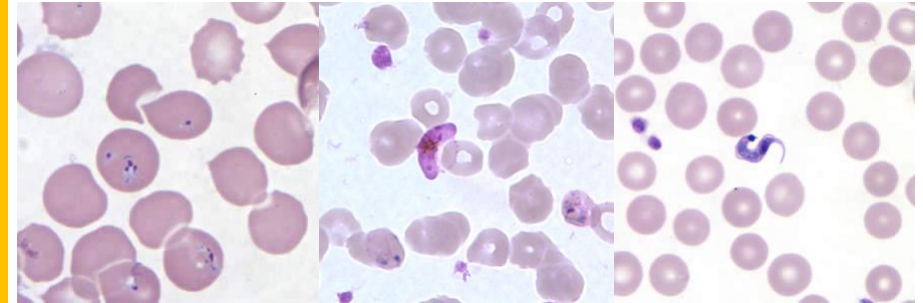


Exercise 4: Improving Network Accuracy

Conclusion

Exercise 2 – Create a network from scratch

- Open `work_DeepLearningFormScratch.mlx` in **02-DeepLearningFromScratchUsingCNN** folder
- If you have an account on Alvis, or running on your own computer – Go to the **breakout room** and finish the exercise
- If you don't have a MATLAB in front of you – follow along in Main Room. Let's do the exercise together!



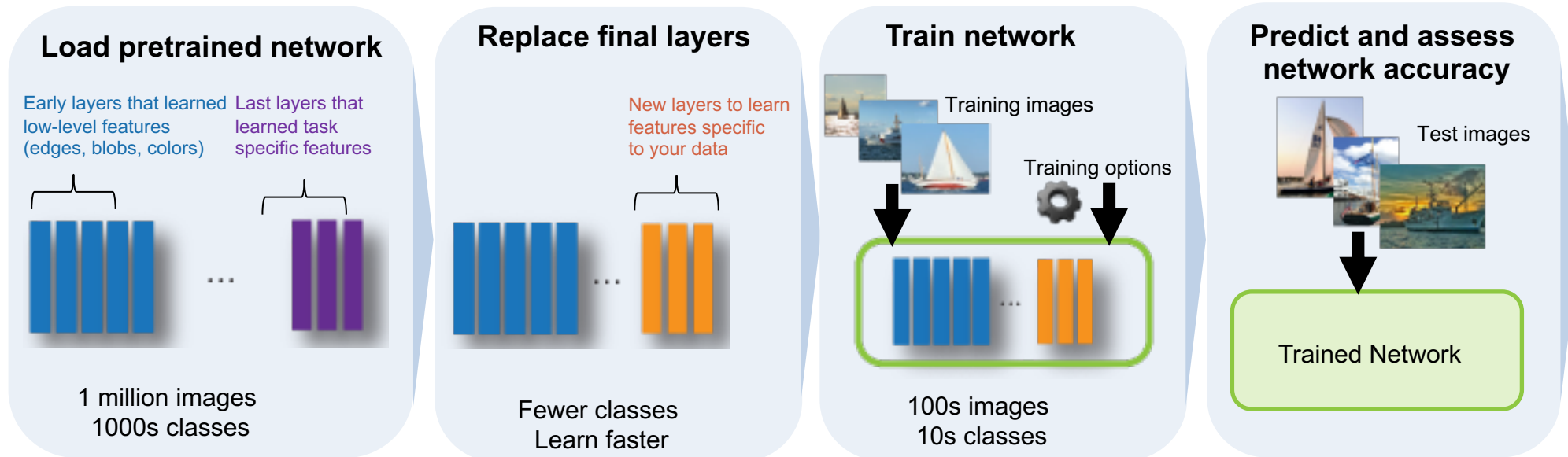
Task:

Create a network that can differentiate between different blood parasites

Workflow:

- Read in data from a folder
- Analyze the data
- Set up a CNN network
- Train network
- Test network

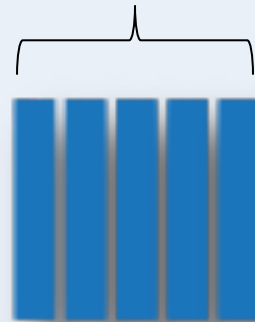
Transfer Learning Workflow



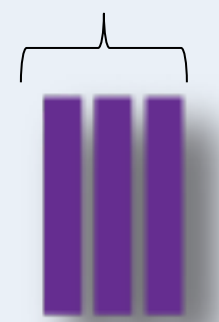
Transfer Learning Workflow – Step 1

Load pretrained network

Early layers learn low-level features (edges, blobs, colors)



Last layers learn task-specific features



...

1 million images
1000s classes

Transfer Learning Workflow – Step 2

Load pretrained network

Early layers that learned low-level features (edges, blobs, colors) Last layers that learned task specific features



1 million images
1000s classes

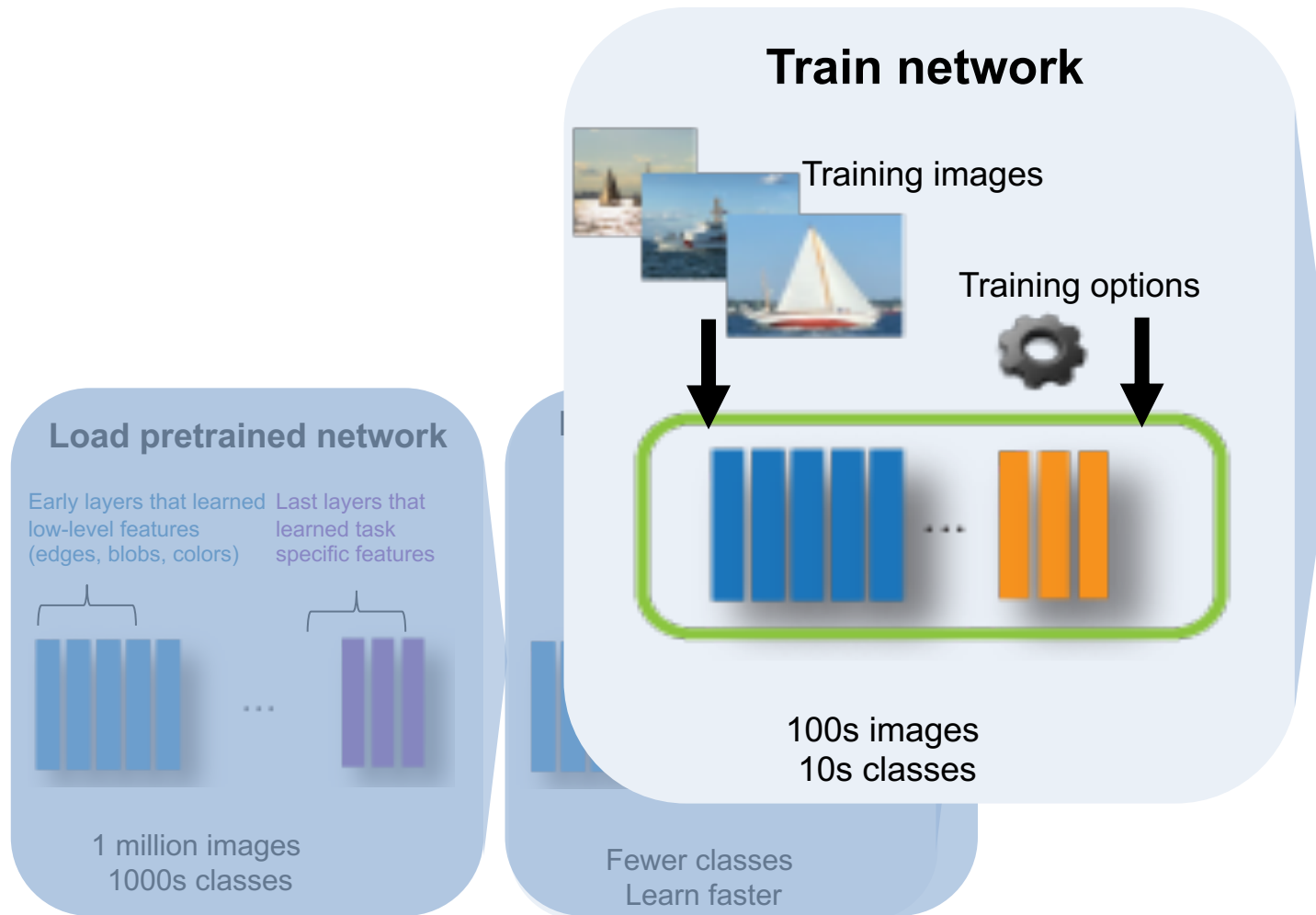
Replace final layers

New layers learn features specific to your data

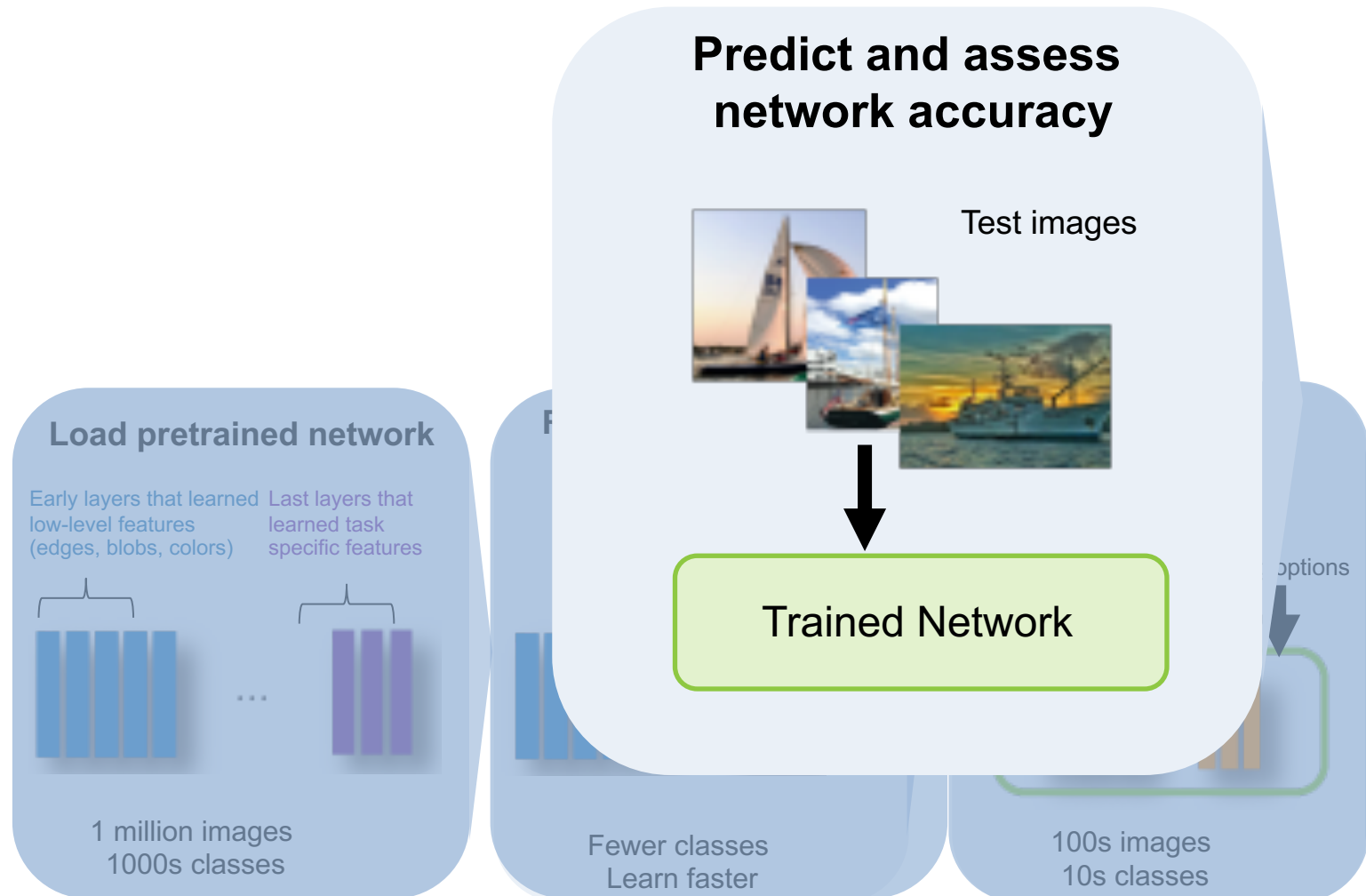


Fewer classes
Learn faster

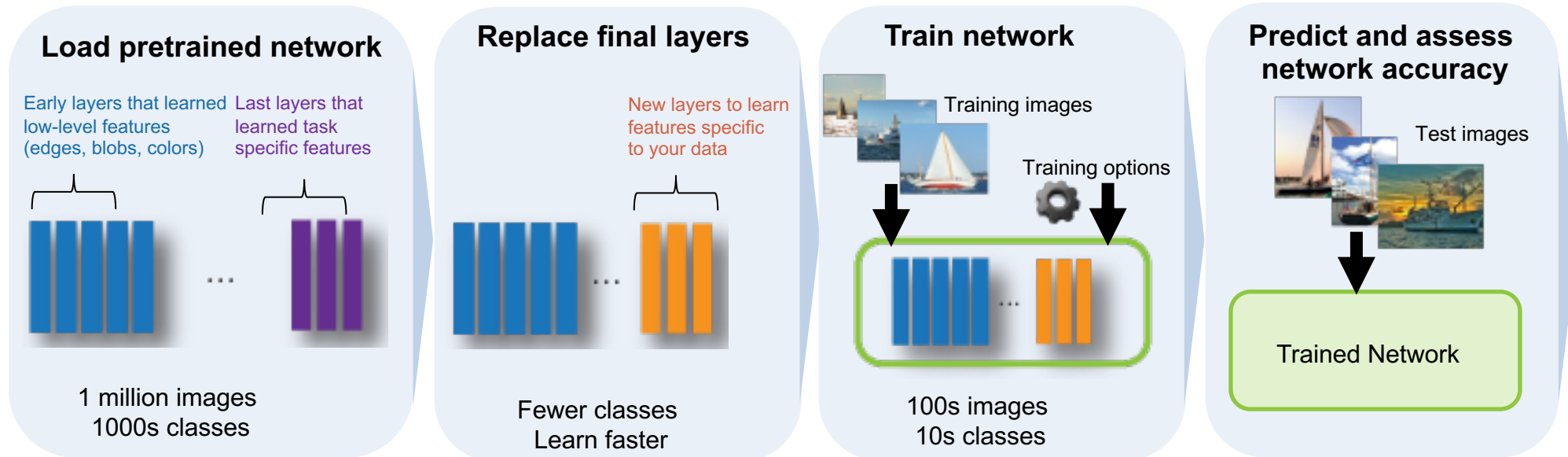
Transfer Learning Workflow – Step 3



Transfer Learning Workflow – Step 4

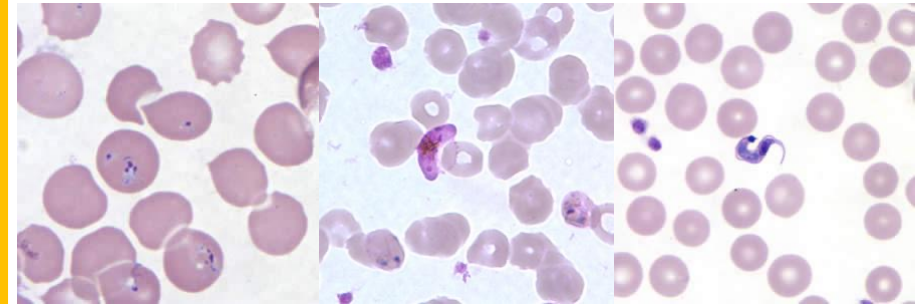


Transfer Learning Workflow



Exercise 3 – Transfer Learning

- Open `work_DeepLearningTransferLearning.mlx` in **03-TransferLearningWithCNN** folder
- If you have an account on Alvis, or running on your own computer – Go to the **breakout room** and finish the exercise
- If you don't have a MATLAB in front of you – follow along in Main Room. Let's do the exercise together!



Task:

Create a network that can differentiate between different blood parasites using a pretrained network

Workflow:

- Read in data from a folder
- Read in pretrained network
- Change layers in the network
- Train network
- Test network

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Exercise 2 & 3: Classify blood smear images

Break



Exercise 4: Improving Network Accuracy

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Exercise 2 & 3: Classify blood smear images

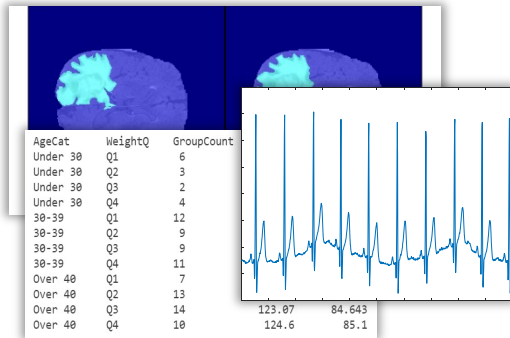
Break



Exercise 4: Improving Network Accuracy

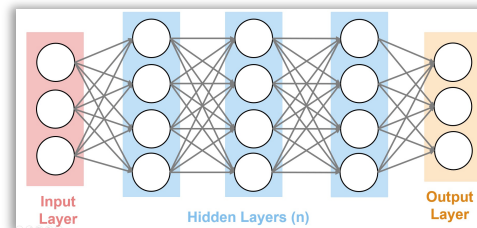
Conclusion

Why is my neural network giving me “incorrect” results?



Data

- Is there **enough** data?
- Do we have the **right** data for the application?
- What about the **quality** of the data?
- Is data **augmentation** in use?



Neural Network Architecture

- What layers are being used and why?
- Are the **appropriate layers** being used for the application?
- Is there **normalization** and/or **regularization**?
- Are you validating during training to check for **overfitting**?

etclearn-rate
batch-size
weights bias
regularization

Parameters and Their Settings

- What parameters are present?
- Were the batch size, learn rate, weights, and other values on “default”?

Exercise 4 – Improving Network Accuracy

- Open `work_DeepLearningImprovingAccuracy.mlx` in **04-ImprovingNetworkAccuracy** folder
- If you have an account on Alvis, or running on your own computer – Go to the **breakout room** and play around with different settings
- If you don't have a MATLAB in front of you – follow along in Main Room. Let's do the exercise together!

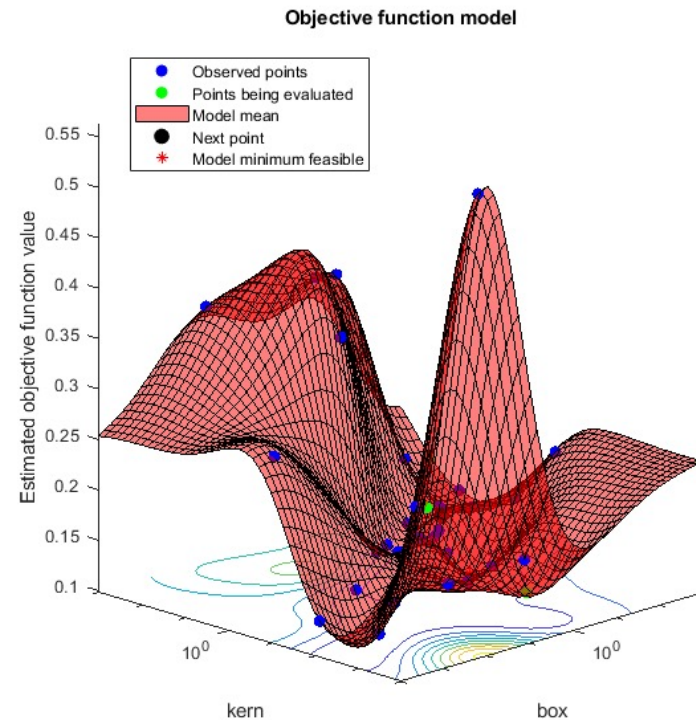
>	'MaxEpochs' – Maximum number of epochs 30 (default) positive integer
>	'MiniBatchSize' – Size of mini-batch 128 (default) positive integer
>	'Shuffle' – Option for data shuffling 'once' (default) 'never' 'every-epoch'
Validation	
>	'ValidationData' – Data to use for validation during training datastore table cell array
>	'ValidationFrequency' – Frequency of network validation 50 (default) positive integer
>	'ValidationPatience' – Patience of validation stopping Inf (default) positive integer
Solver Options	
>	'InitialLearnRate' – Initial learning rate 0.001 0.01 positive scalar
>	'LearnRateSchedule' – Option for dropping learning rate during training 'none' (default) 'piecewise'
>	'LearnRateDropPeriod' – Number of epochs for dropping the learning rate 10 (default) positive integer

Hyperparameter Tuning in Deep Learning

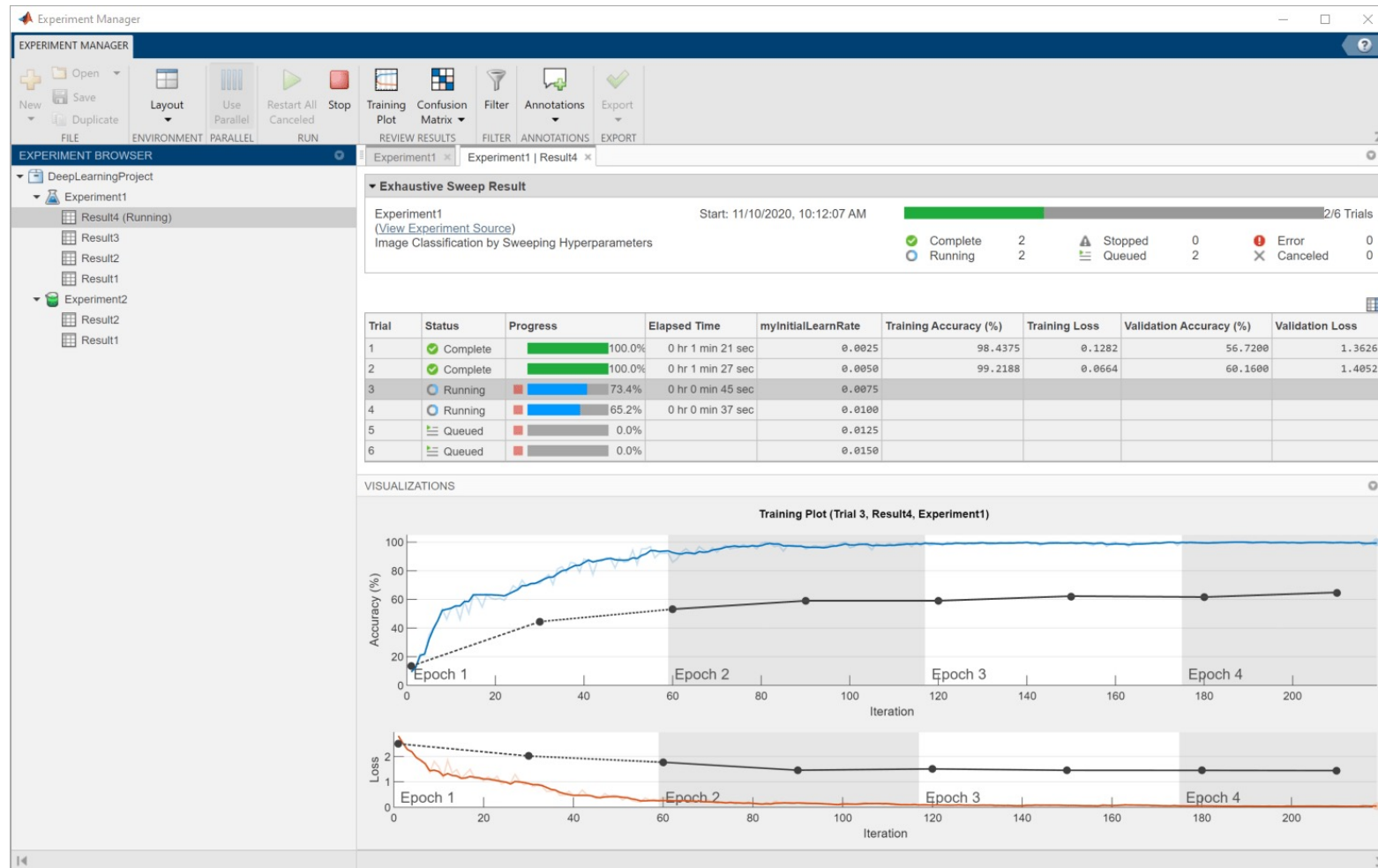
“The challenge with hyperparameters is that there are no magic number that works everywhere. The best numbers depend on each task and each dataset.”

[Source](#)

Hyperparameter tuning using Bayesian Optimization



Experiment Manager



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Exercise 4: Improving Network Accuracy

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Home Exercises

Try LSTM models for time series data:

- Classify heart status based on ECG signals
- work_ClassifyECGSignals.mlx
- <https://sft.mathworks.com/human.aspx?r=245144200&arg06=776492647-15764732a8bb11675e2f947023798fa0&arg12=filelist>
- Username: AppleHill Password: fjr5n8nf

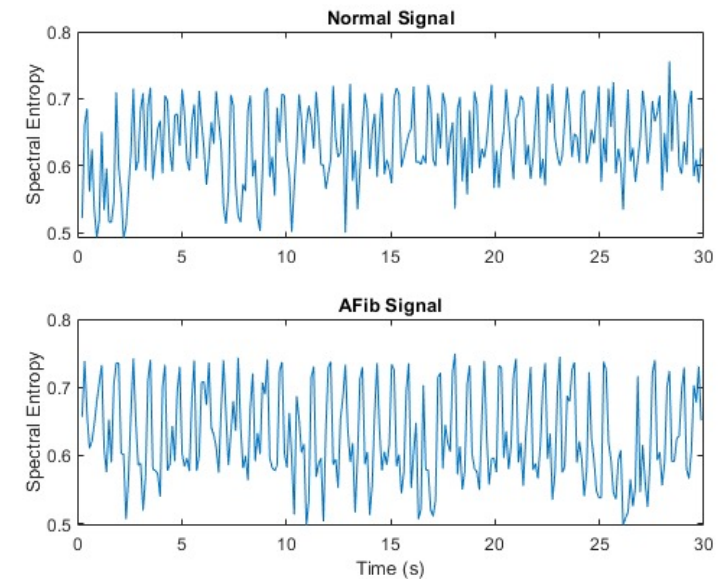
Try a larger dataset for GPU speed up on Alvis:

- Classify pixels in a Semantic Segmentation problem
- <https://se.mathworks.com/help/vision/ug/semantic-segmentation-using-deep-learning.html>
- ~5 hours training time on a low performing GPU. What can you get on the Tesla v100??

The spectral entropy measures how spiky flat the spectrum of a signal is. A signal with a spiky spectrum has a high spectral entropy. The pentropy function estimates the spectral entropy based on a power spectral density (PSD) spectrogram which results in 255 time windows for a signal of 9000 samples. The 255-long time series of spectral entropy values can be visualized as a line plot.

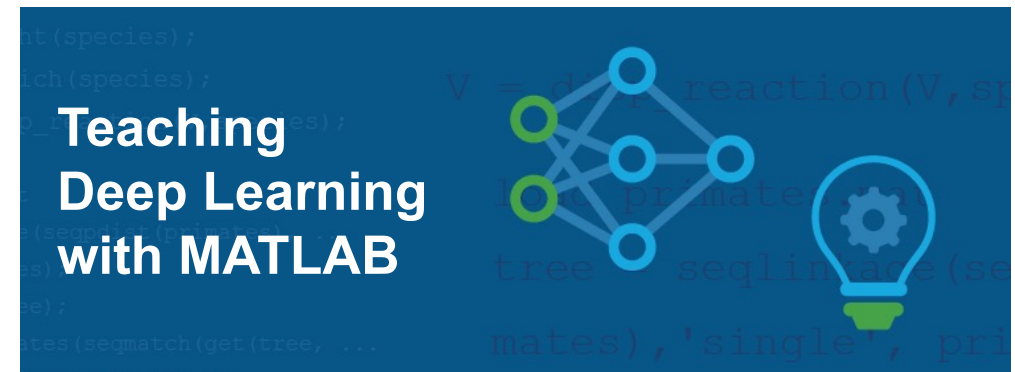
Visualize the spectral entropy for each type of signal.

```
[pentropyA,tA2] = pentropy(aFib,fs);  
[pentropyN,tN2] = pentropy(normal,fs);  
  
plotPentropy(tN2,pentropyN,tA2,pentropyA);
```



Further Learning and Teaching

- [Deep Learning Onramp](#)
 - 2 hr online tutorial
- [Deep Learning with MATLAB self-paced course](#)
 - >14 hours interactive tutorial
- Deep Learning Workshop
 - 3 hr hands on session
 - Contact us to schedule
- [Deep Learning Training](#)
 - 16 hr in depth course
 - Online or Instructor Lead
- [Teaching Deep Learning with MATLAB](#)
 - Curriculum support



Good Luck with the home exercises!