

Convolutional Neural Network

Data Workshop #7

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DATA
WORKSHOP

Disclaimer

Data Workshop [*all time*] focuses on the **intuition** and **practical** tips.

For a formal treatment, see something else^{}.*

^{*} papers or classical machine learning books

Environment

[github.com/databricks/prerequisite](https://github.com/databricks/databricks-prereq)
[github.com/databricks/environment](https://github.com/databricks/databricks-environment)

[github.com/databricks/cnn](https://github.com/databricks/databricks-cnn)

Packages

github.com/datworkshop/prerequisite

```
$ python run.py
seaborn-0.7.0 - OK
xgboost-0.4 - OK
matplotlib-1.5.1 - OK
IPython-4.1.2 - OK
numpy-1.11.0 - OK
pandas-0.18.0 - OK
sklearn-0.17.1 - OK
```

```
=====
All right, you are ready to go on Data Workshop!
```

```
$ python run.py
seaborn-0.6 should be upgraded to seaborn-0.7
xgboost-0.4 - OK
matplotlib-1.5.1 - OK
IPython-4.1.2 - OK
numpy-1.11.0 - OK
pandas-0.18.0 - OK
sklearn-0.17.1 - OK
```

```
=====
RECOMENDATION (without upgrade some needed features could be missing)
pip install --upgrade seaborn
```

```
$ python run.py
seaborn-0.7.0 - OK
xgboost - missing
matplotlib-1.5.1 - OK
IPython-4.1.2 - OK
numpy-1.11.0 - OK
pandas-0.18.0 - OK
sklearn-0.17.1 - OK
```

```
=====
REQUIRED
Please install those packages before Data Workshop: xgboost
pip install xgboost
More info how to install xgboost: http://xgboost.readthedocs.org/en/latest/build.html
```

jupyter notebook



```
$ jupyter notebook
[I 22:17:17.650 NotebookApp] The port 8888 is already in use, trying another random port.
[I 22:17:17.650 NotebookApp] The port 8889 is already in use, trying another random port.
[I 22:17:17.651 NotebookApp] The port 8890 is already in use, trying another random port.
[I 22:17:17.651 NotebookApp] The port 8891 is already in use, trying another random port.
[I 22:17:17.657 NotebookApp] Serving notebooks from local directory: /Users/vova/src/github/dataworkshop/titanic/vladimir/tmp
[I 22:17:17.657 NotebookApp] 0 active kernels
[I 22:17:17.657 NotebookApp] The IPython Notebook is running at: http://localhost:8892/
[I 22:17:17.657 NotebookApp] Use Control-C to stop this server and shut down all kernels (twice to skip confirmation).
```



Files Running Clusters

Select items to perform actions on them.



Notebook list empty.



Text File

Folder

Terminal

Notebooks

Haskell

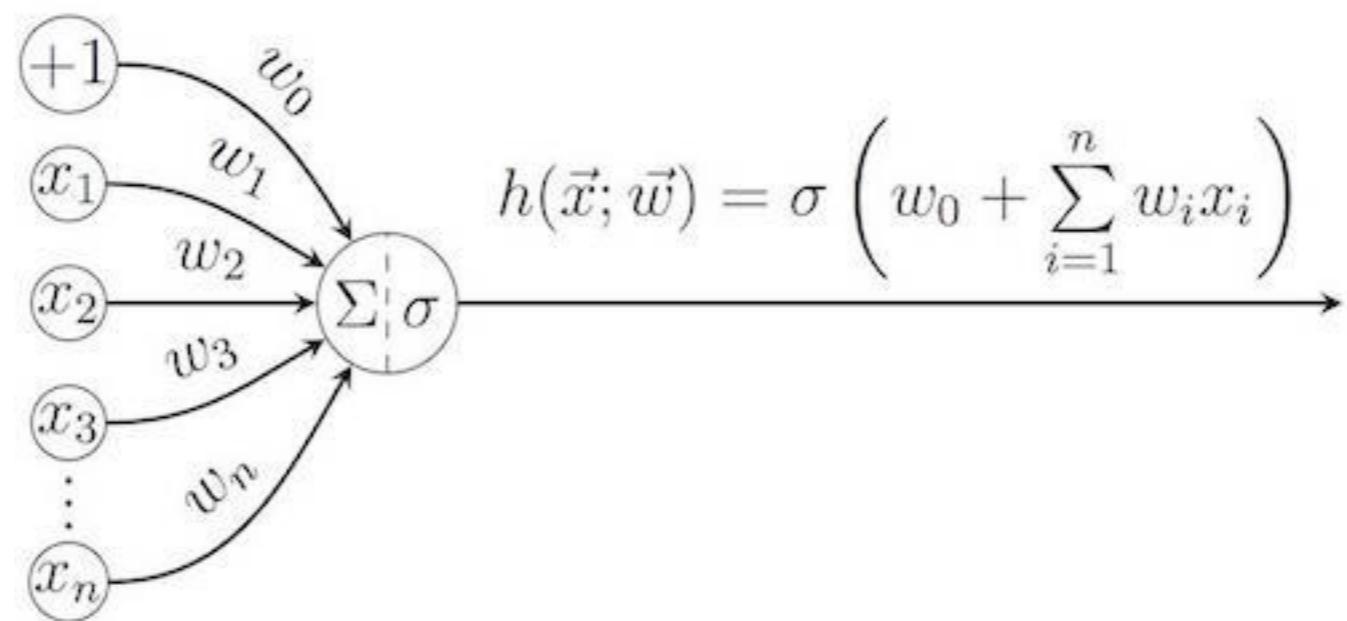
Julia 0.3.8

Python 2



Motivation

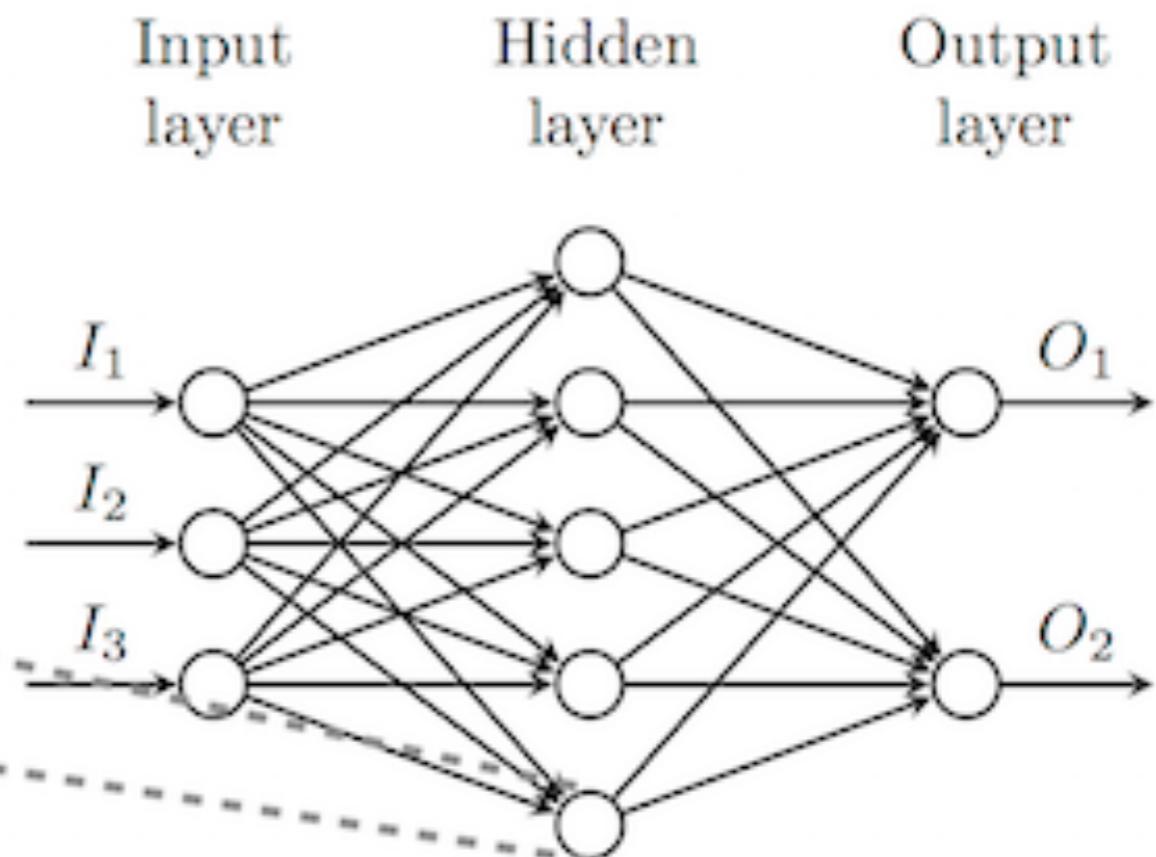
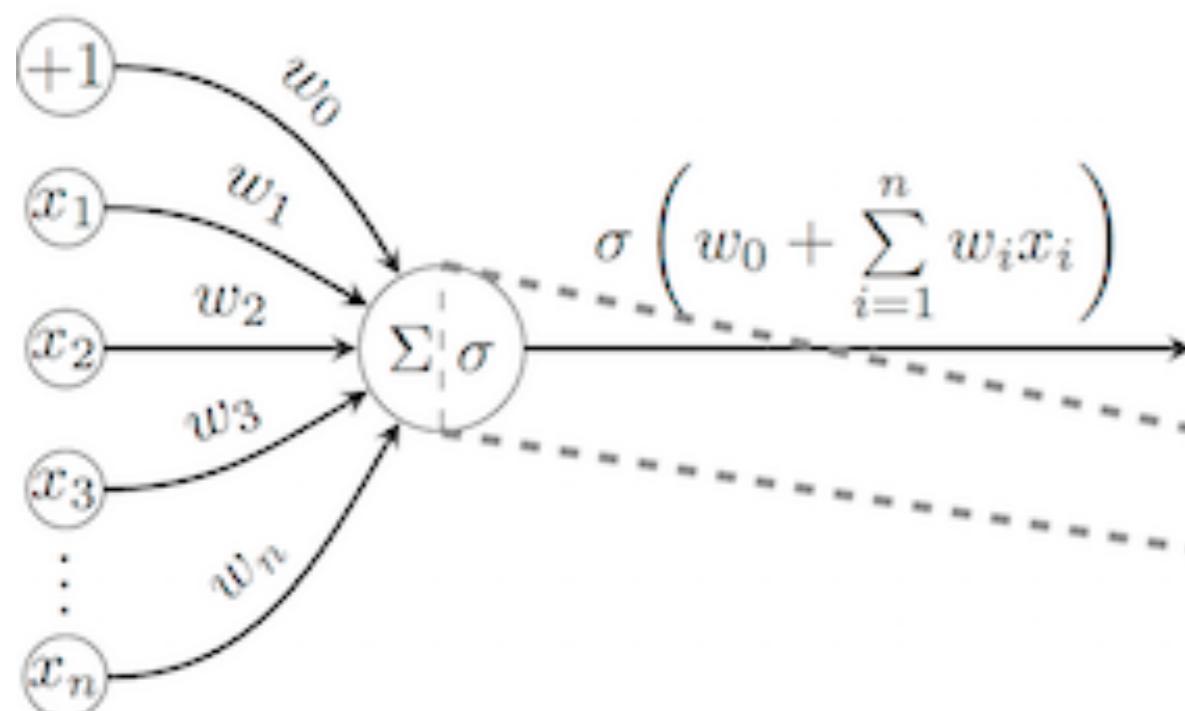
Neuron



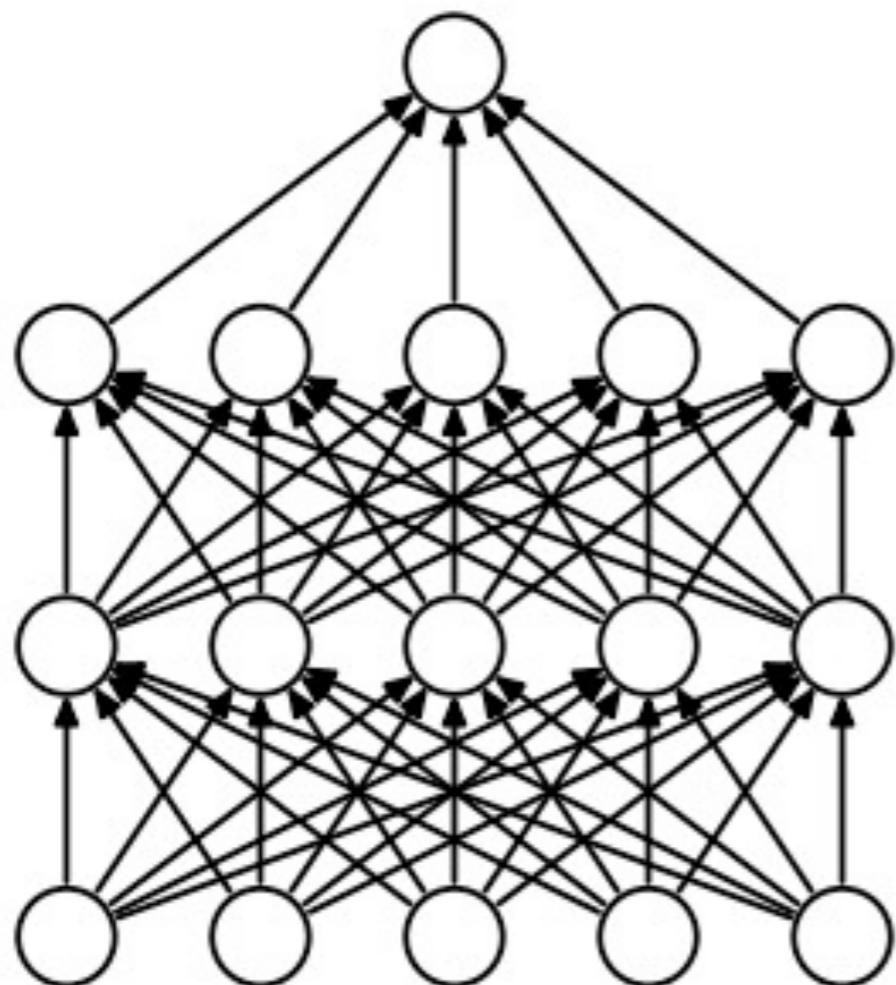
Activation function

- Tanh
- ReLU
- PRELU
- Sigmoid
- SoftMax (normalized exponential function)

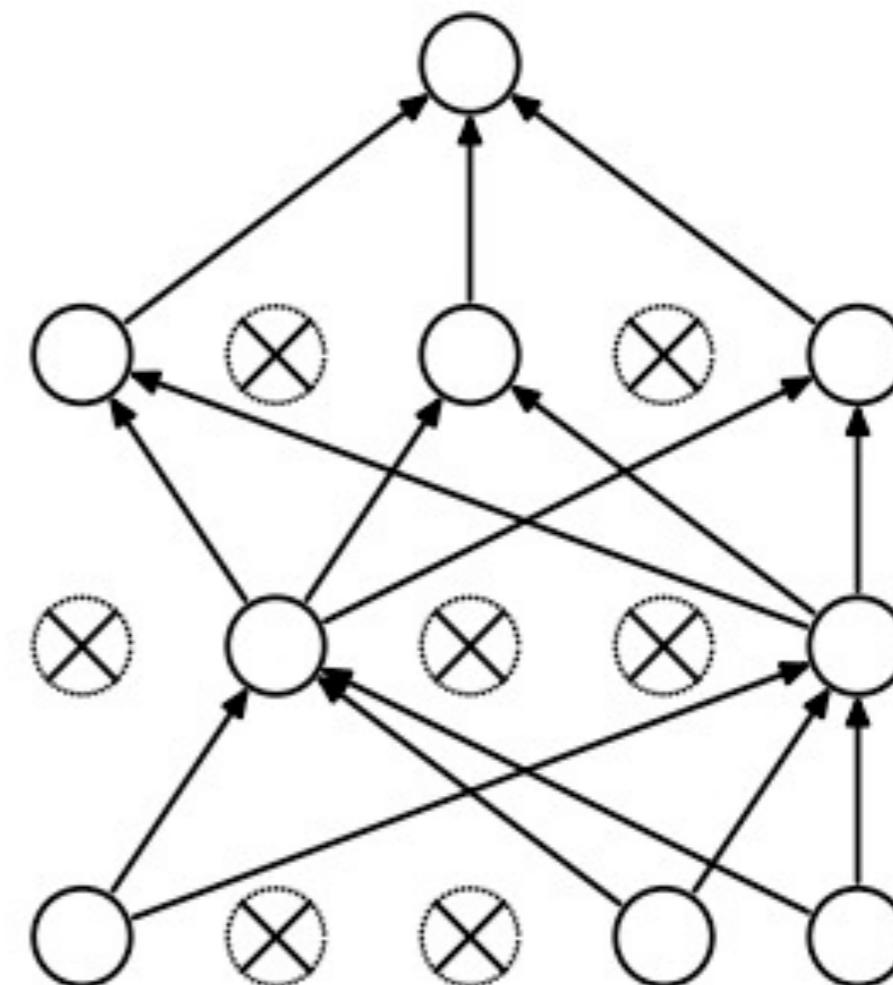
(Artificial) Neural Network



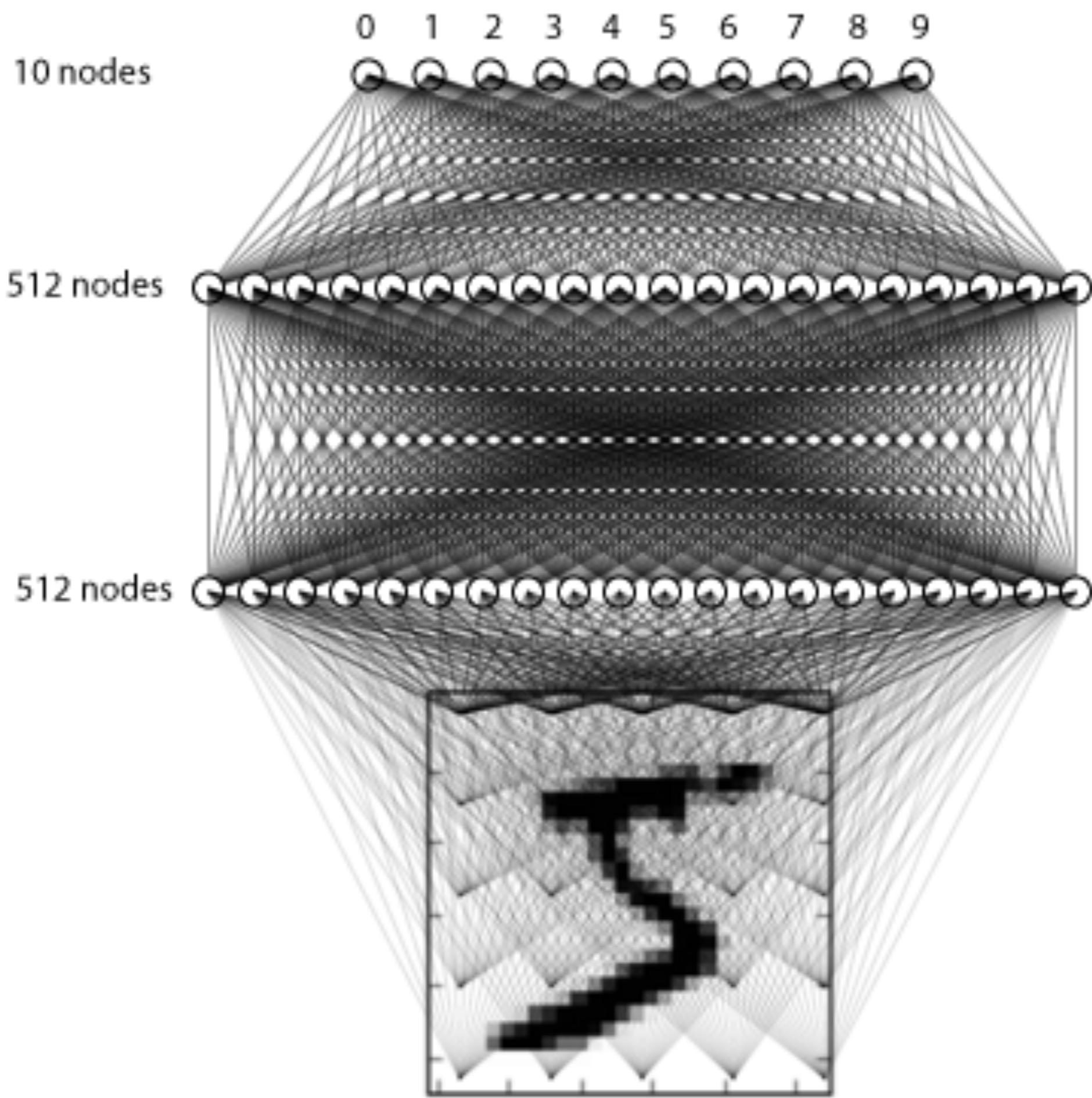
Dropout



(a) Standard Neural Net



(b) After applying dropout.



MNIST

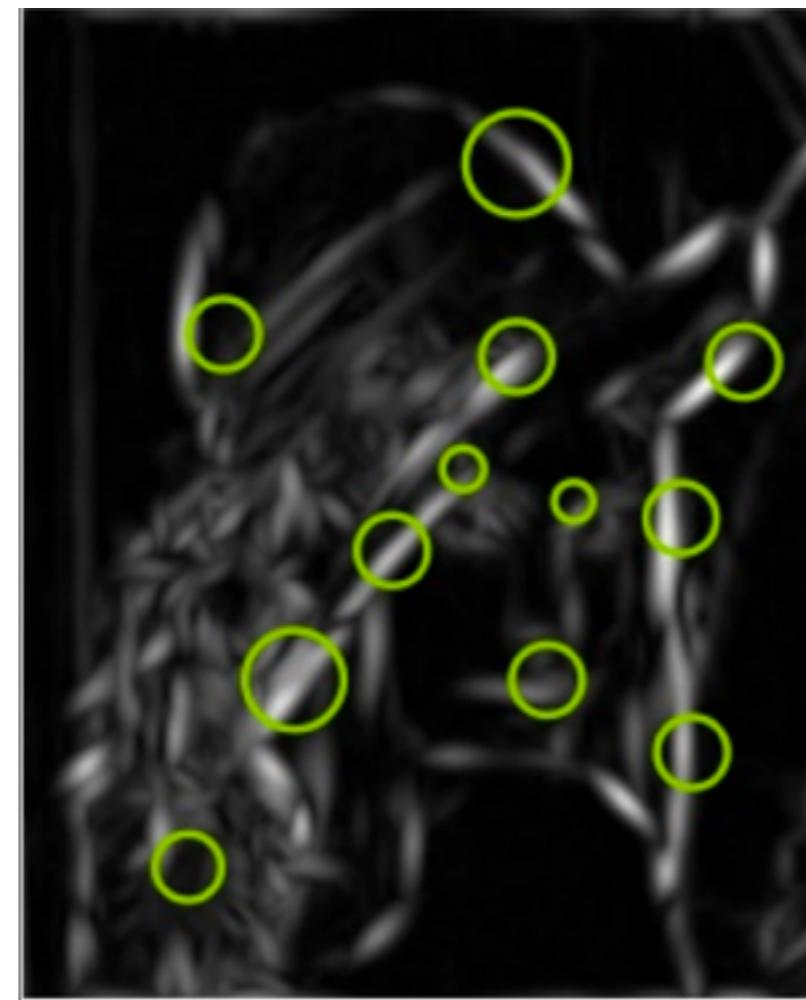
0 0 0 0 0 0 0 0 0 0
1 1 1 1 1 1 1 1 1 1
2 2 2 2 2 2 2 2 2 2
3 3 3 3 3 3 3 3 3 3
4 4 4 4 4 4 4 4 4 4
5 5 5 5 5 5 5 5 5 5
6 6 6 6 6 6 6 6 6 6
7 7 7 7 7 7 7 7 7 7
8 8 8 8 8 8 8 8 8 8
9 9 9 9 9 9 9 9 9 9

Computer Vision

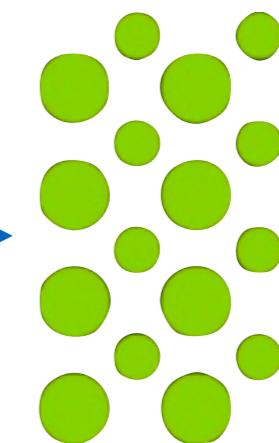
Original



Simplified

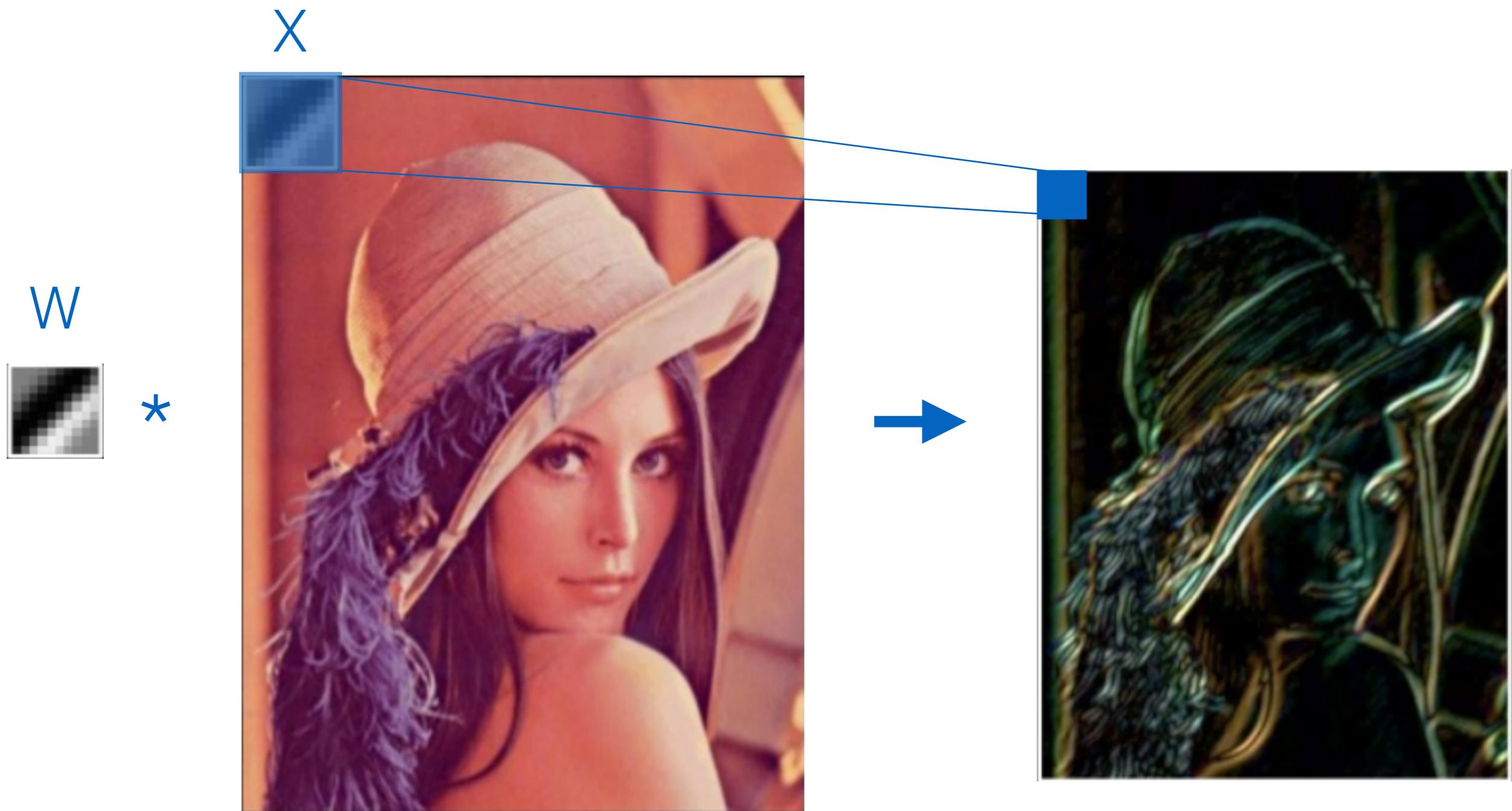


Features



→ “Lenna”

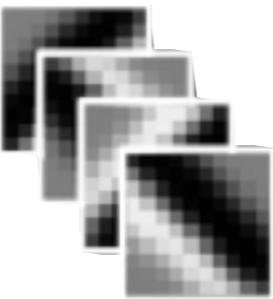
Feature Map



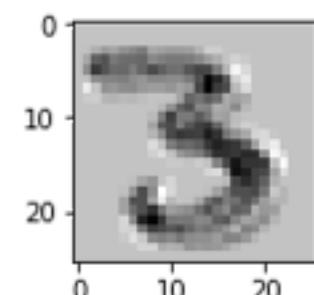
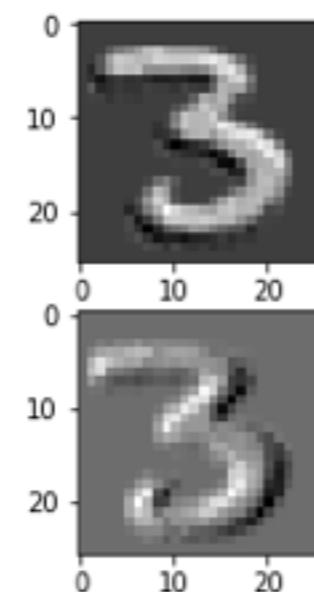
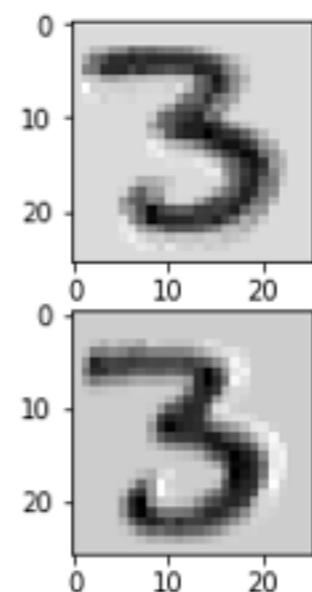
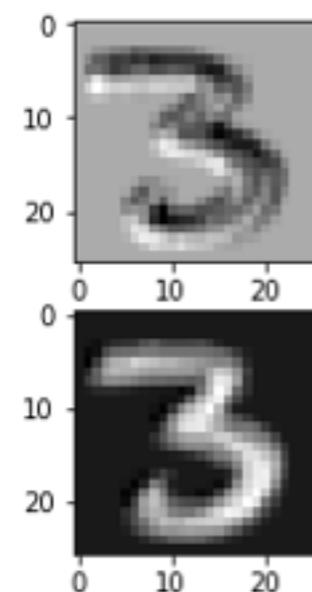
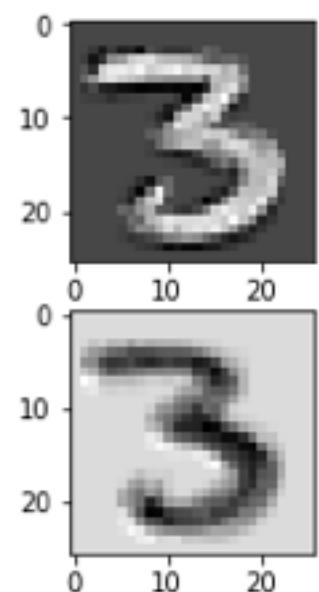
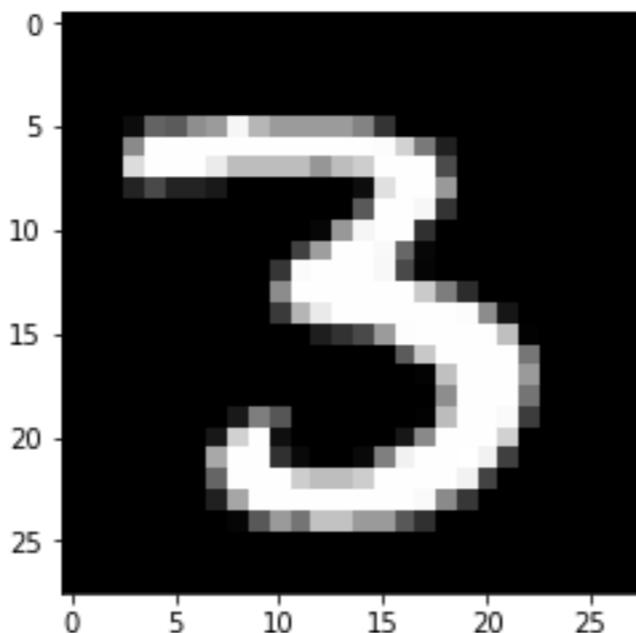
Feature Maps



*



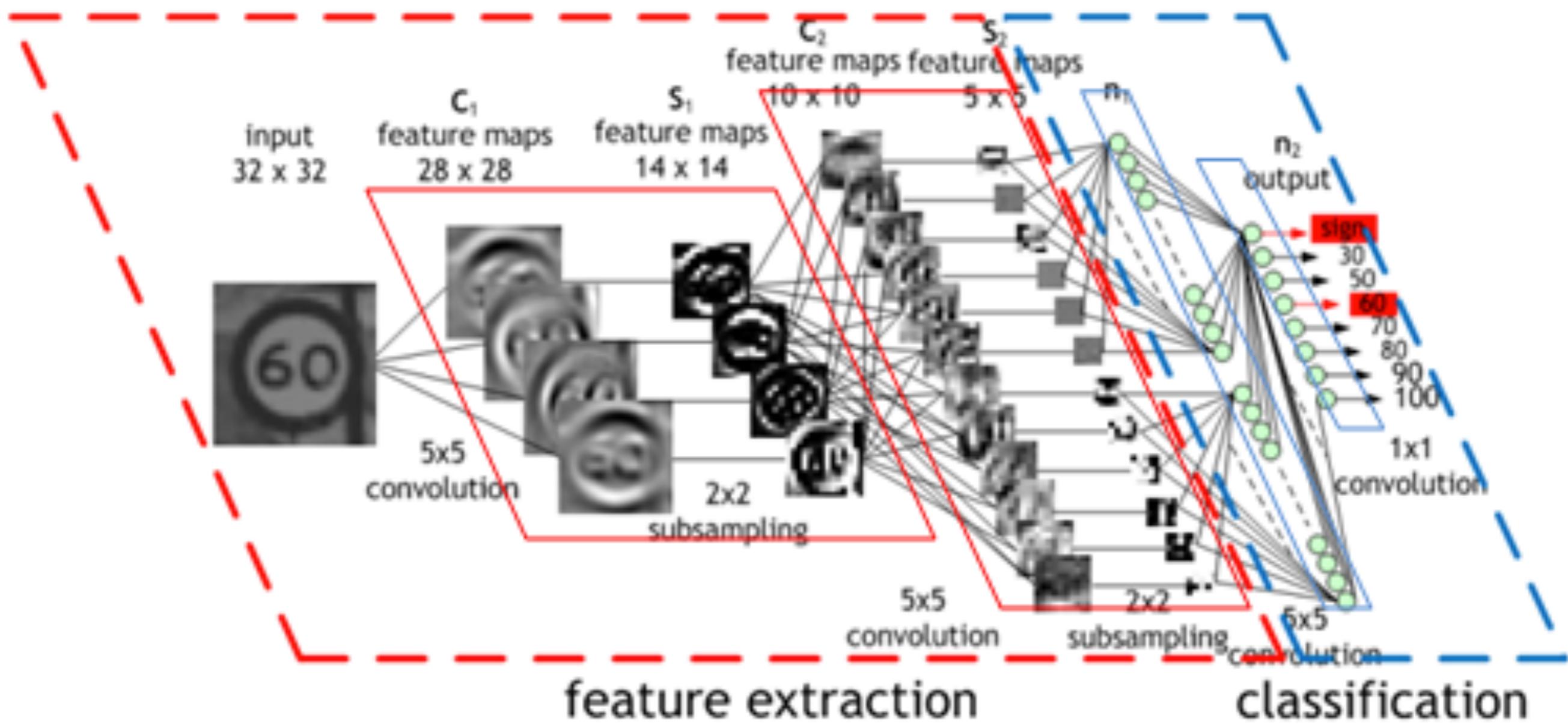
Feature Maps

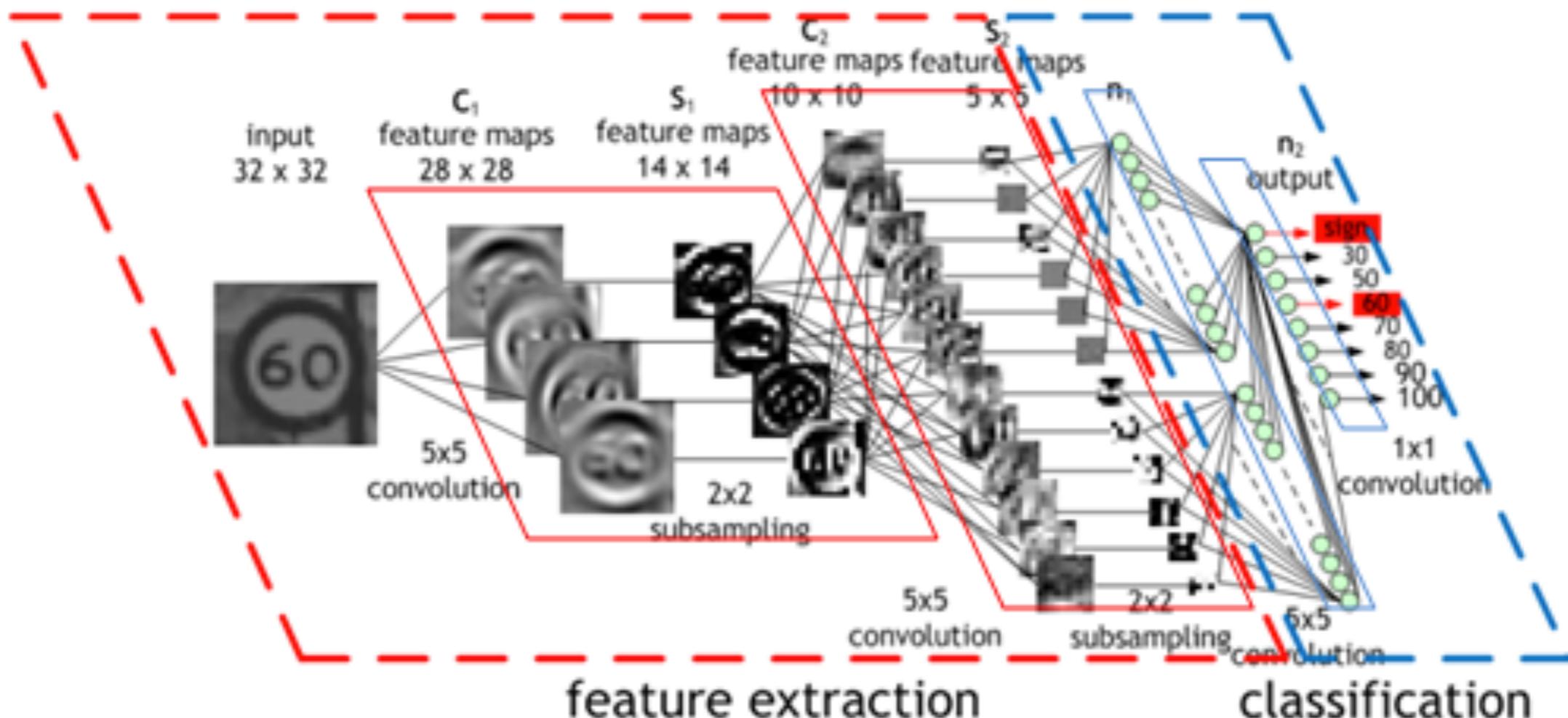


Pooling == resize



Convolutional Neural Networks





```

model = Sequential()

model.add(Conv2D(4, kernel_size=(5, 5), activation='relu', input_shape=(1, 32, 32)))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Conv2D(12, (5, 5), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dense(512, activation='relu'))
model.add(Dense(100, activation='softmax'))

model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=[ 'accuracy' ])

```

Curse of machine learning



Reuse knowledge

Transfer Learning

Lifelong Learning

Gradual Learning

Curriculum Learning

Compositional Learning

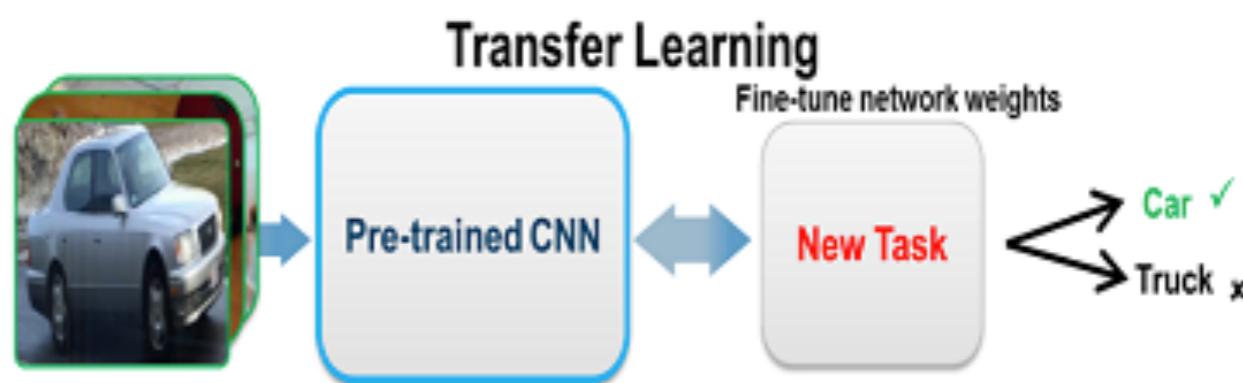
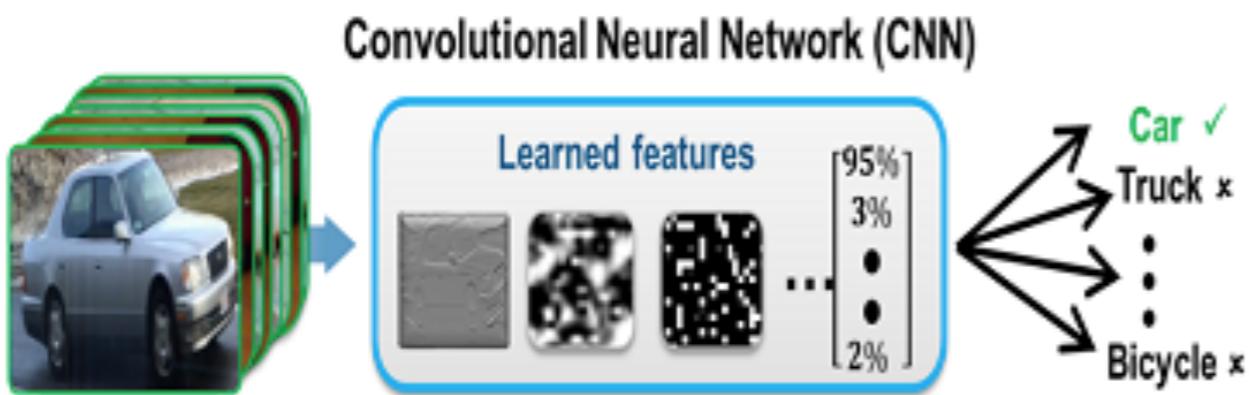
Continual Learning

Sequential Learning

Never-Ending Learning



Transfer Learning & Fine-tuning

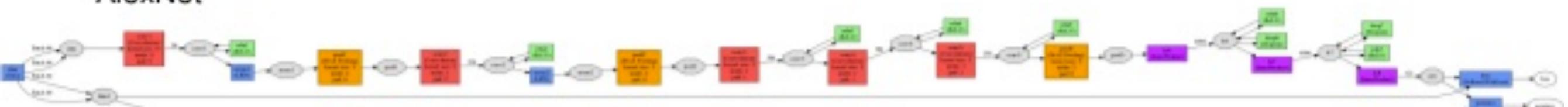


Training data	1000s to millions of labeled images
Computation	Compute intensive
Training Time	Days to Weeks for real problems
Model accuracy	High (can over fit to small datasets)

Training data	100s to 1000s of labeled images (small)
Computation	Moderate computation
Training Time	Seconds to minutes
Model accuracy	Good, depends on the pre-trained CNN model

Known Topology - ImageNet

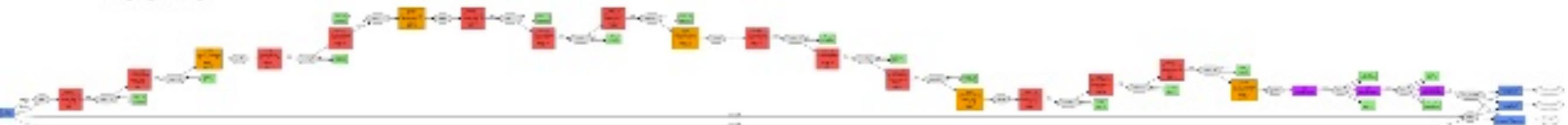
AlexNet



GoogleNet



VGG-16



ResNet-50





```
from keras.applications.resnet50 import ResNet50
from keras.preprocessing import image
from keras.applications.resnet50 import preprocess_input, decode_predictions
import numpy as np

model = ResNet50(weights='imagenet')

img = image.load_img('cat.jpg', target_size=(224, 224))
x = np.expand_dims(image.img_to_array(img), axis=0)
x = preprocess_input(x)

preds = model.predict(x)
print('Predicted:', decode_predictions(preds, top=3)[0])
```



Tabby 47.4%



Lynx 39.9%



Egyptian cat 4.8%



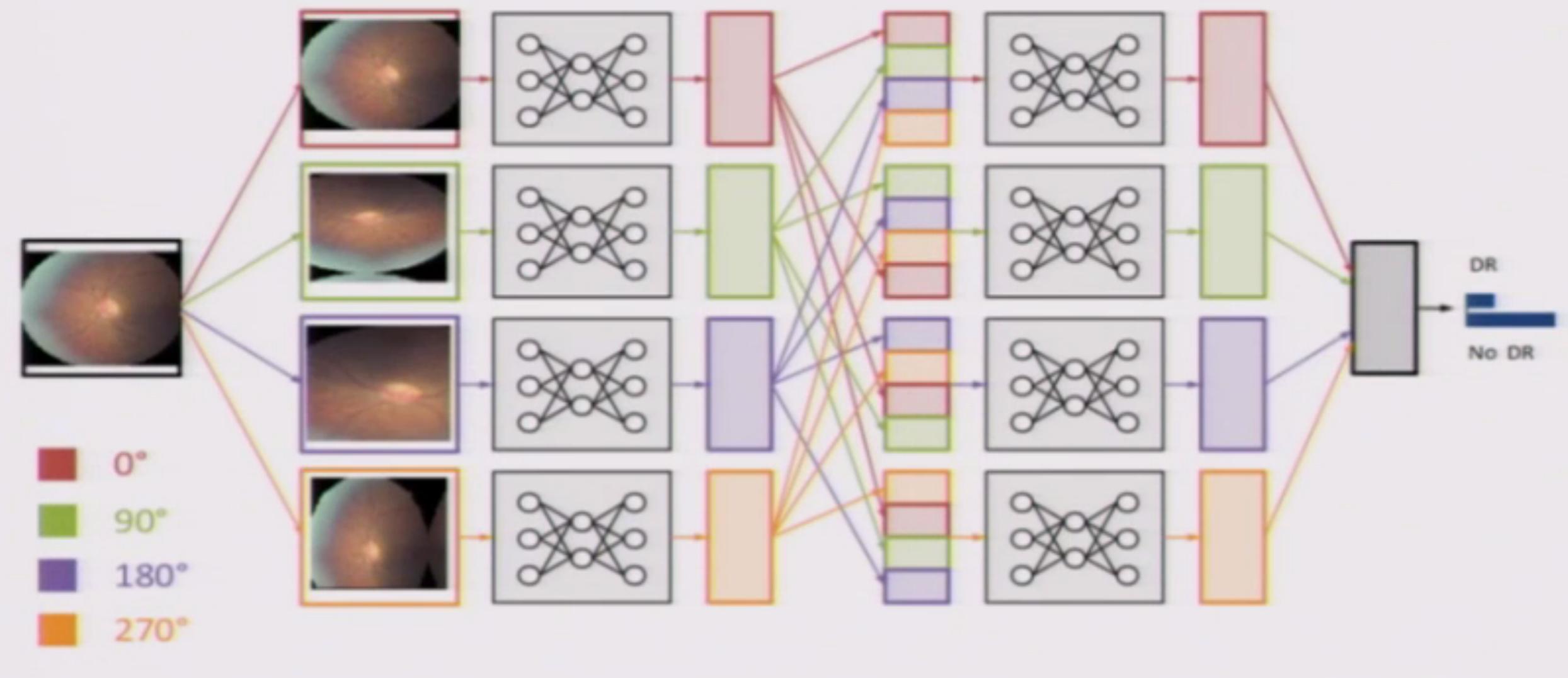


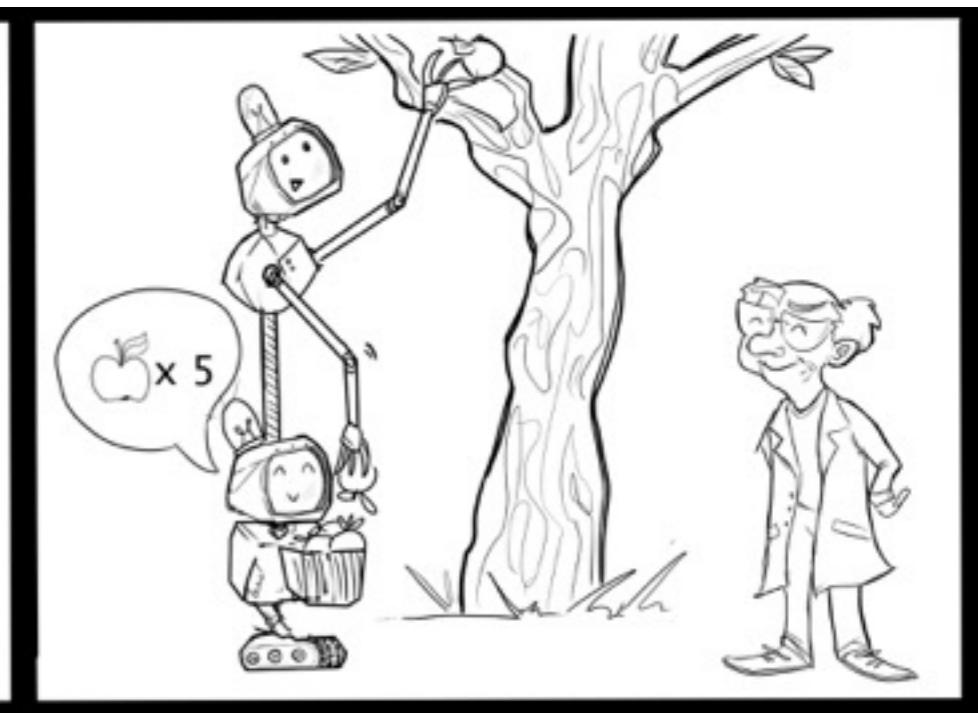
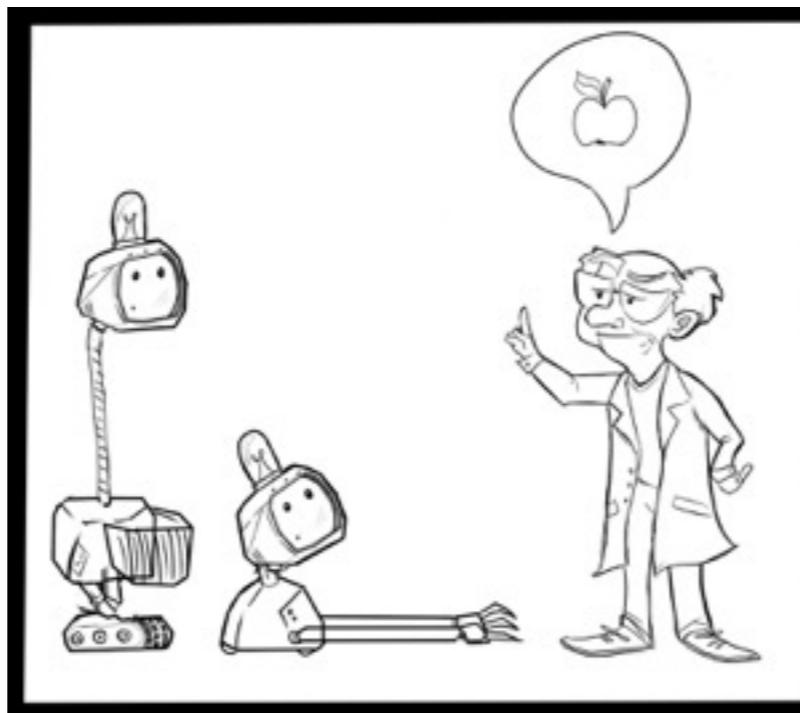
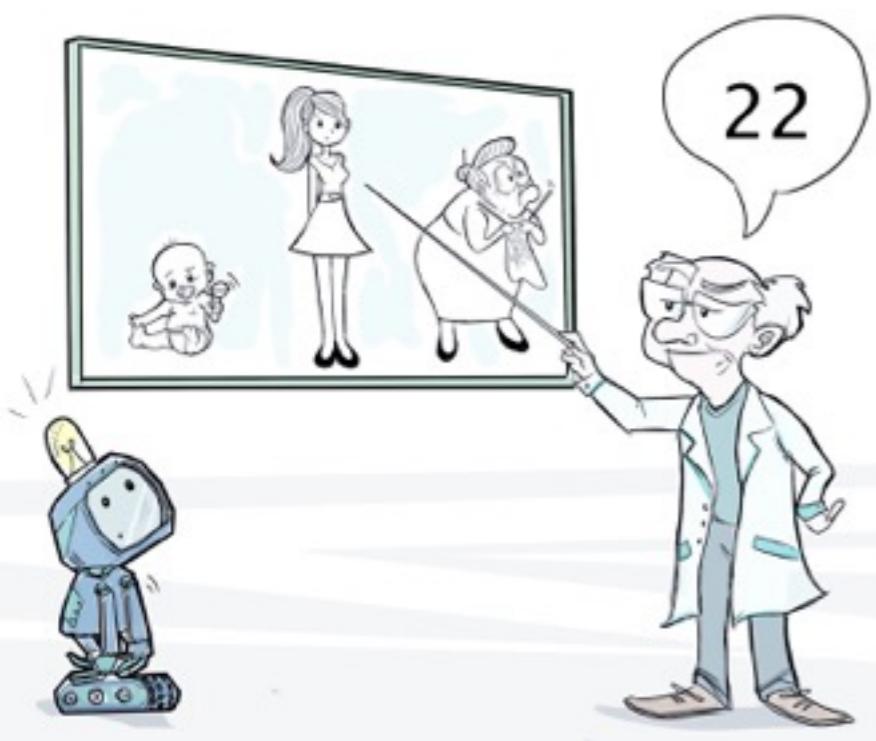
IMAGE CLASSIFICATION	MEAN ACCURACY
Our fine-tuned DCNN	0.96
Feature Based SVM	0.82

BOOK

Humanly about
machine learning



goo.gl/tbqina



Three things

if you can remember only three...

- CNN boost computer vision
- CNN this is also about NLP, language translation...
- CNN required a lot of resources (GPU)

Thank you



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