

KAIST Summer Session 2018

Module 3. Deep Learning with PyTorch

Convolutional Neural Network

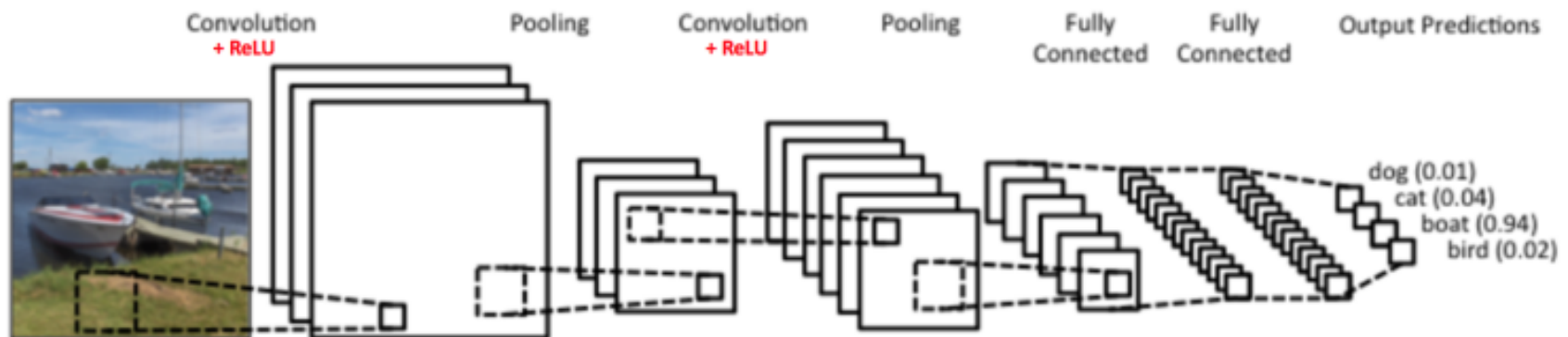
KAIST College of Business

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20 August, 2018

Review: What is CNN?

- A typical example of CNN



- Inputs (2-dimension \times channels) \rightarrow [Convolution \rightarrow ReLu \rightarrow Pooling] \rightarrow ...
 \rightarrow [Convolution \rightarrow ReLu \rightarrow Pooling] \rightarrow Fully connected layer
 \rightarrow Output prediction (Multi-class classification)

Review: What is CNN?

- Convolution

1	1	1	0	0
0	1	1	1	0
0	0	1	1	1
0	0	1	1	0
0	1	1	0	0

Image data

1	0	1
0	1	0
1	0	1

Convolution filter

1	1	1	0	0
0	1	1	1	0
0	0	1	1	1
0	0	1	1	0
0	1	1	0	0

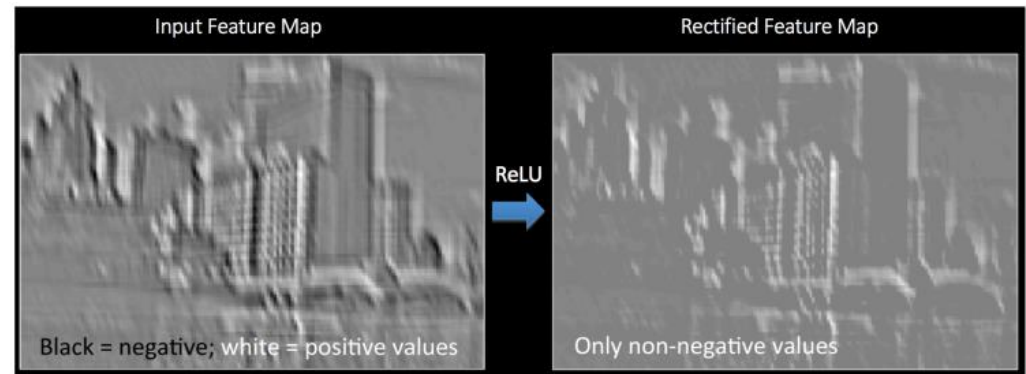
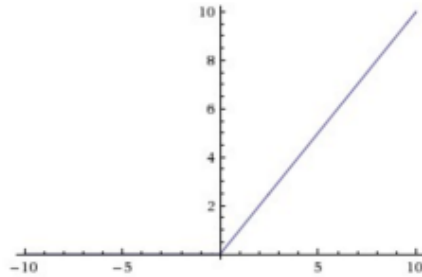
Image

4		

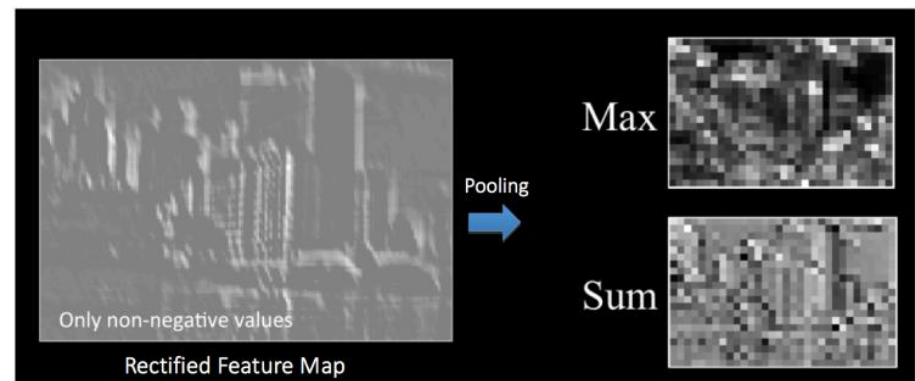
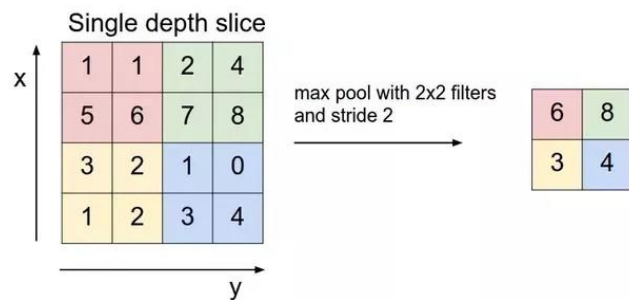
Convolved
Feature

Review: What is CNN?

- ReLU

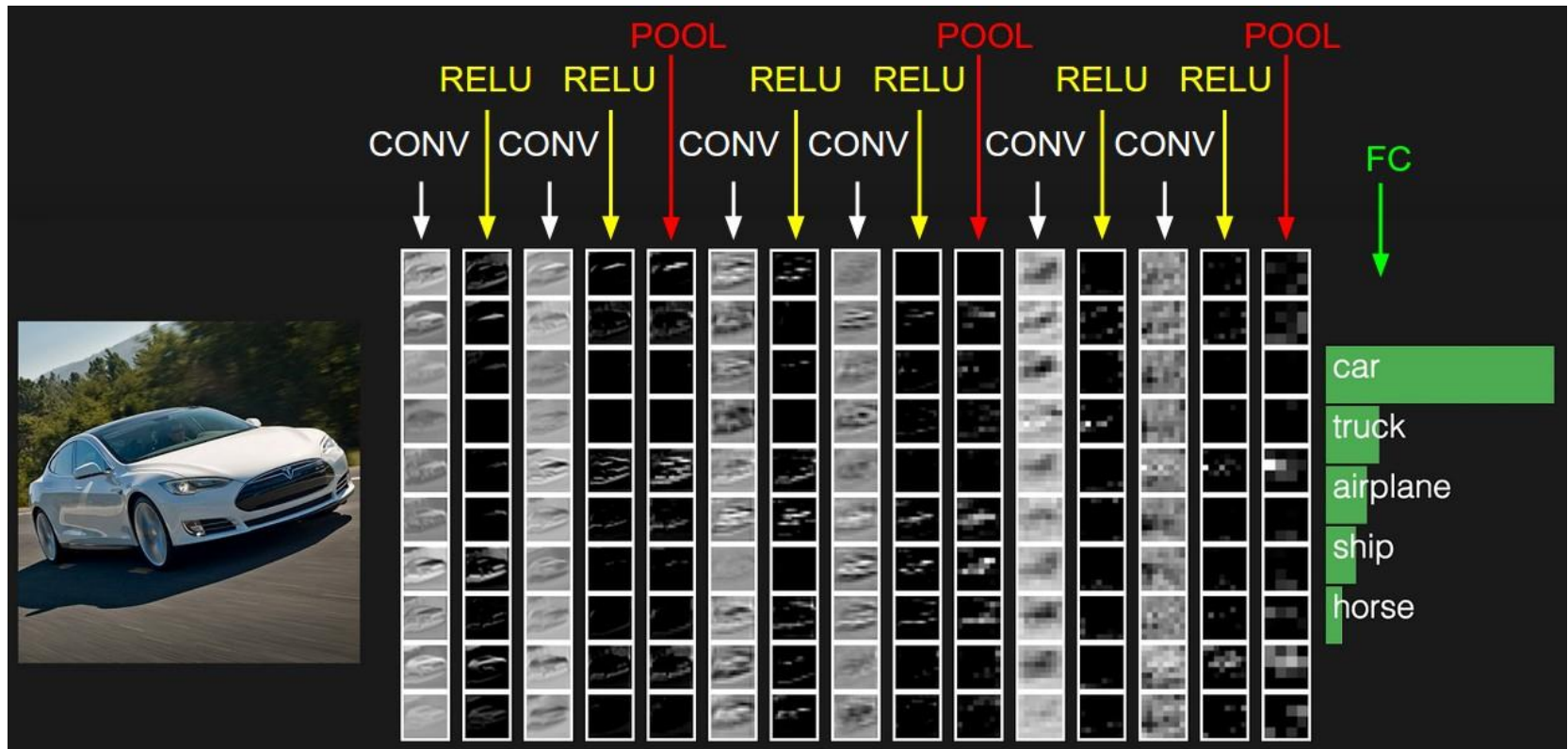


- Max pooling



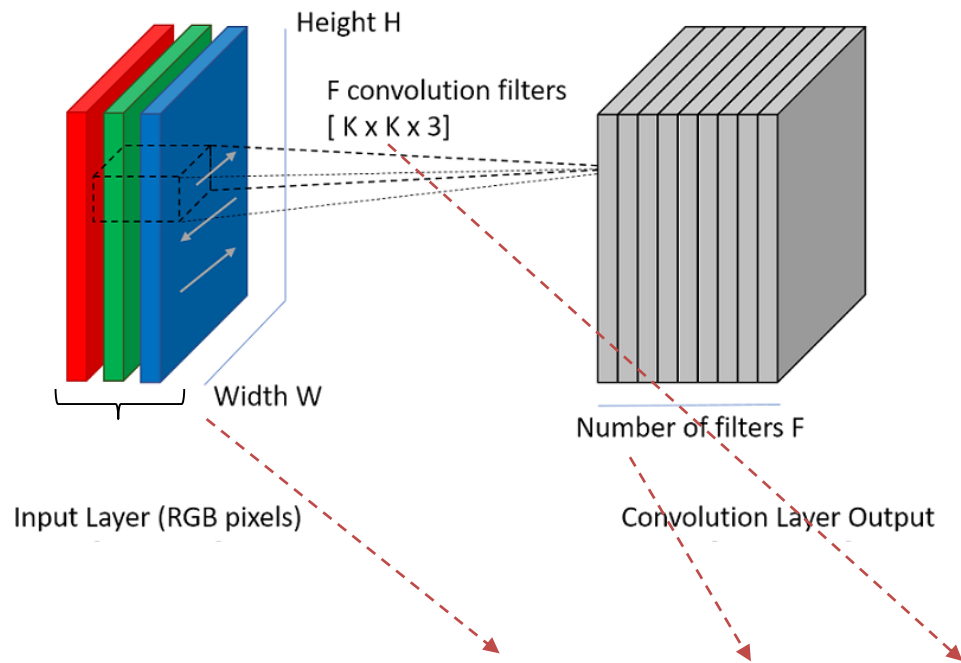
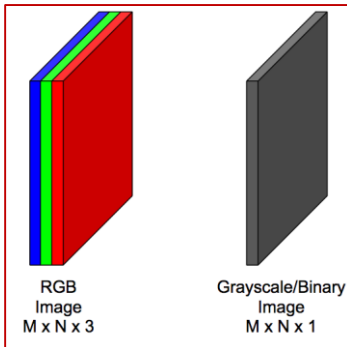
Review: What is CNN?

- Hierarchical feature representation (multiple convolution layers)



CNN in PyTorch

- Convolution



```
self.conv1 = torch.nn.Conv2d(num_of_channels, num_of_kernels, kernel_size)
```

```
self.conv2 = torch.nn.Conv2d(num_of_channels, num_of_kernels, kernel_size)
```

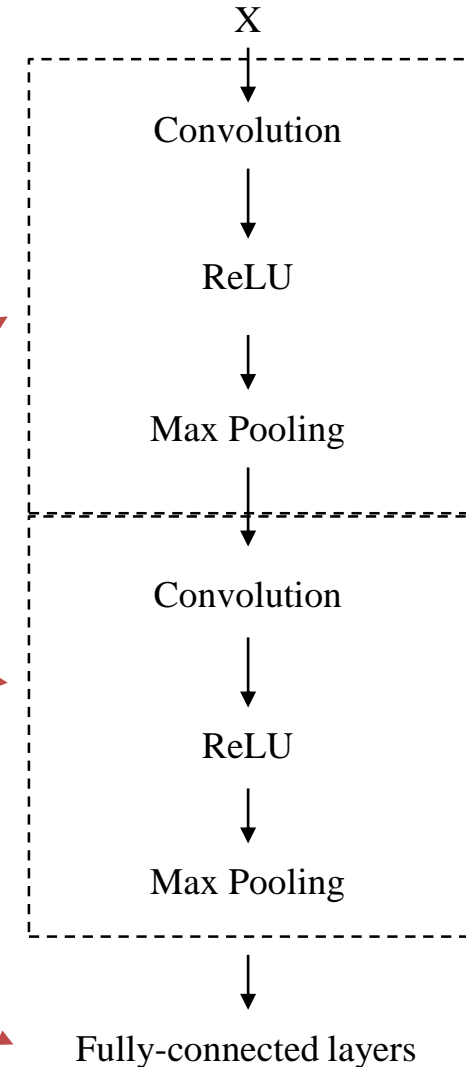
CNN in PyTorch

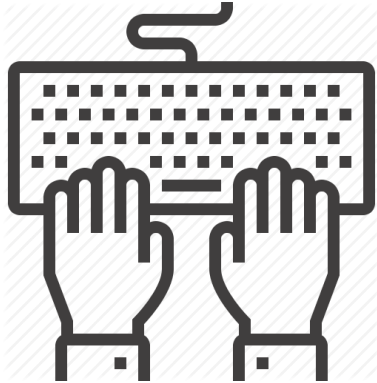
```
import torch.nn as nn
import torch.nn.functional as F

class Net(nn.Module):
    def __init__(self):
        super(Net, self).__init__()
        self.conv1 = nn.Conv2d(3, 6, 5)
        self.pool = nn.MaxPool2d(2, 2)
        self.conv2 = nn.Conv2d(6, 16, 5)
        self.fc1 = nn.Linear(16 * 5 * 5, 120)
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)

    def forward(self, x):
        x = self.pool(F.relu(self.conv1(x)))
        x = self.pool(F.relu(self.conv2(x)))
        x = x.view(-1, 16 * 5 * 5)
        x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = self.fc3(x)
        return x

net = Net()
```



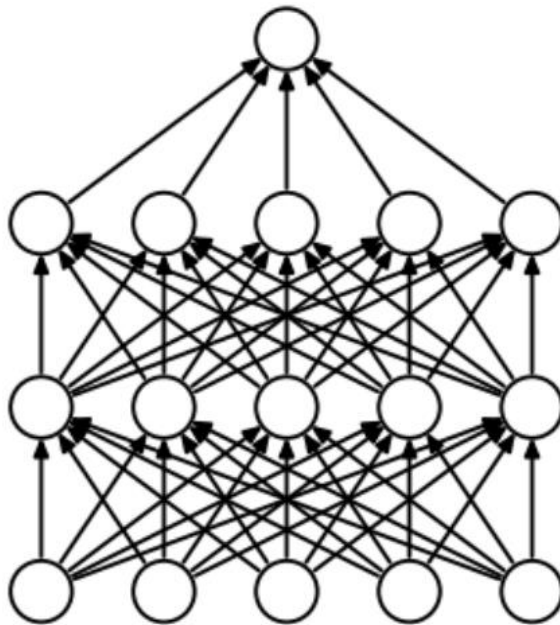


MNIST Classifier using CNN

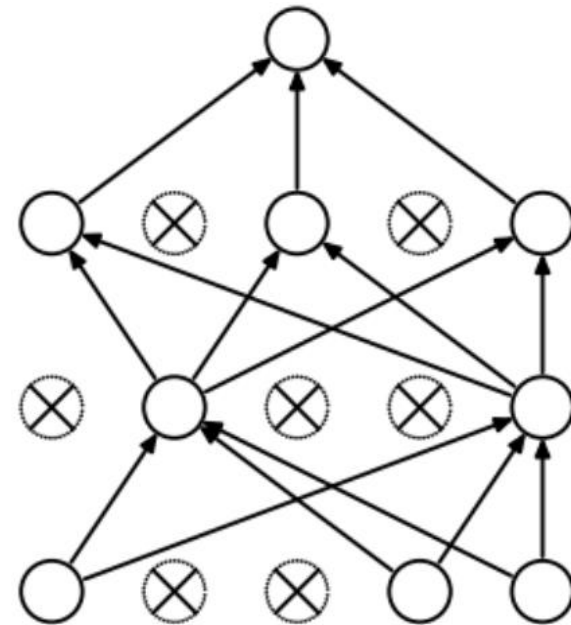
*M3.5 Convolutional Neural
Network_MNIST.ipynb*

Improving Performance (1) Dropout

- Dropout avoids overfitting of neural networks



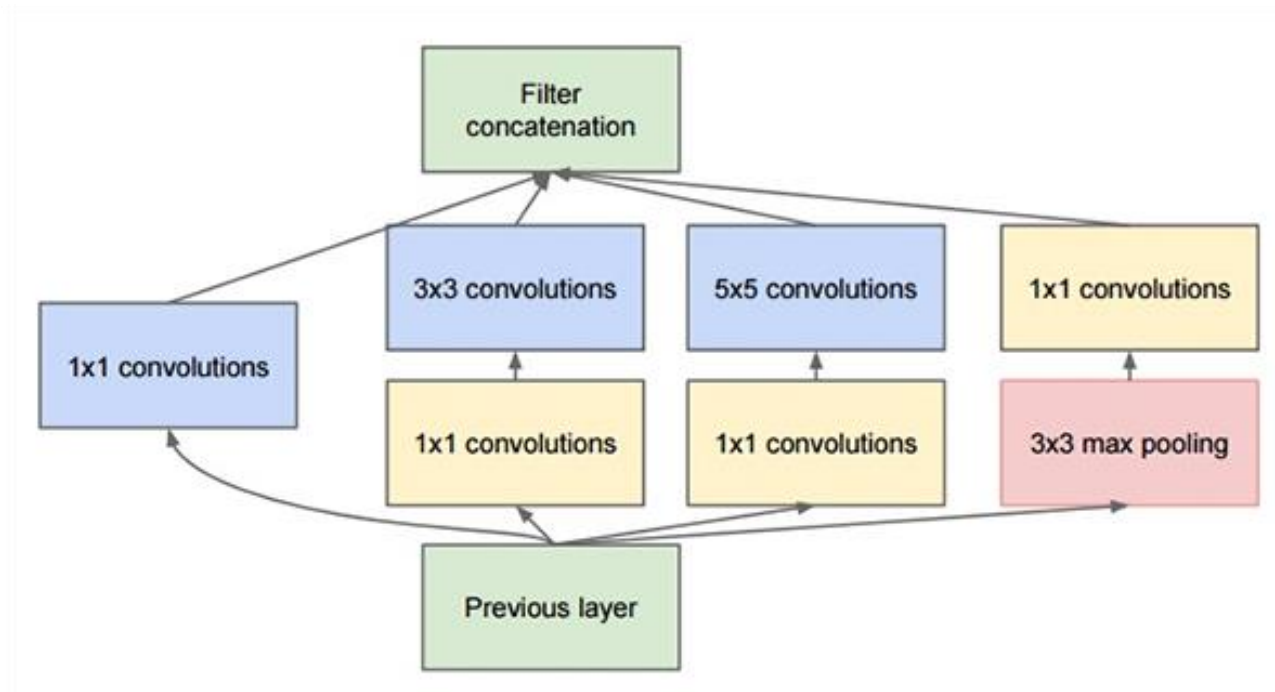
(a) Standard Neural Net



(b) After applying dropout.

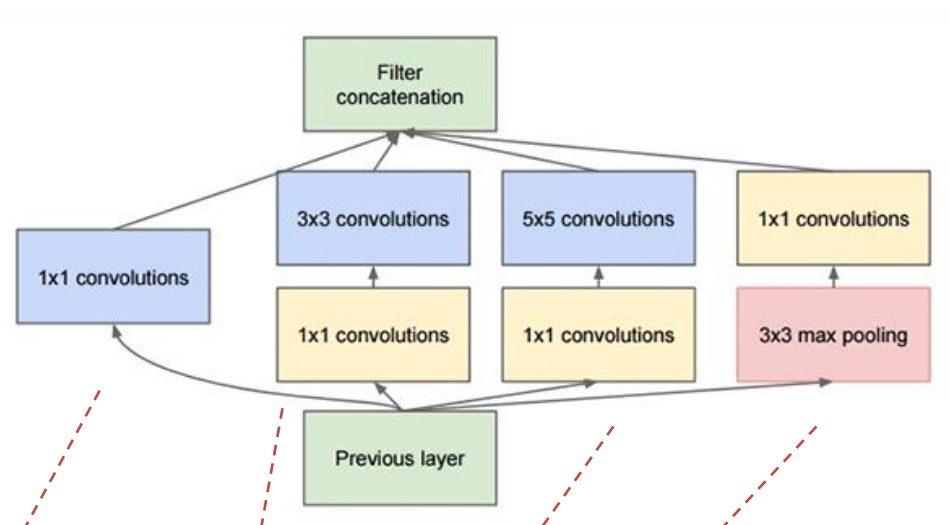
Improving Performance (2) Inception

- CNN can have filters with multiple sizes operating on the same level.
 - CNN with inception tends to be more efficient and has better performance (Szegedy et al. 2015)



Szegedy, C., Liu, W., Jia, Y., Sermanet, P., Reed, S., Anguelov, D., Erhan, D., Vanhoucke, V. and Rabinovich, A., 2015. Going Deeper with Convolutions. *In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*.

Improving Performance (2) Inception



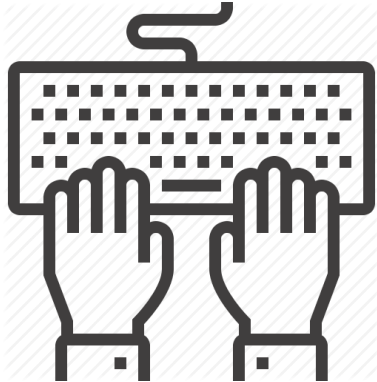
```
class Inception(nn.Module):

    def __init__(self, in_channels):
        super(Inception, self).__init__()
        self.branch1x1 = nn.Conv2d(in_channels, 16, kernel_size=1)

        self.branch5x5_1 = nn.Conv2d(in_channels, 16, kernel_size=1)
        self.branch5x5_2 = nn.Conv2d(16, 24, kernel_size=5, padding=2)

        self.branch3x3dbl_1 = nn.Conv2d(in_channels, 16, kernel_size=1)
        self.branch3x3dbl_2 = nn.Conv2d(16, 24, kernel_size=3, padding=1)

        self.pool = nn.MaxPool2d(kernel_size=3, stride=1, padding=1)
        self.branch_pool = nn.Conv2d(in_channels, 24, kernel_size=1)
```



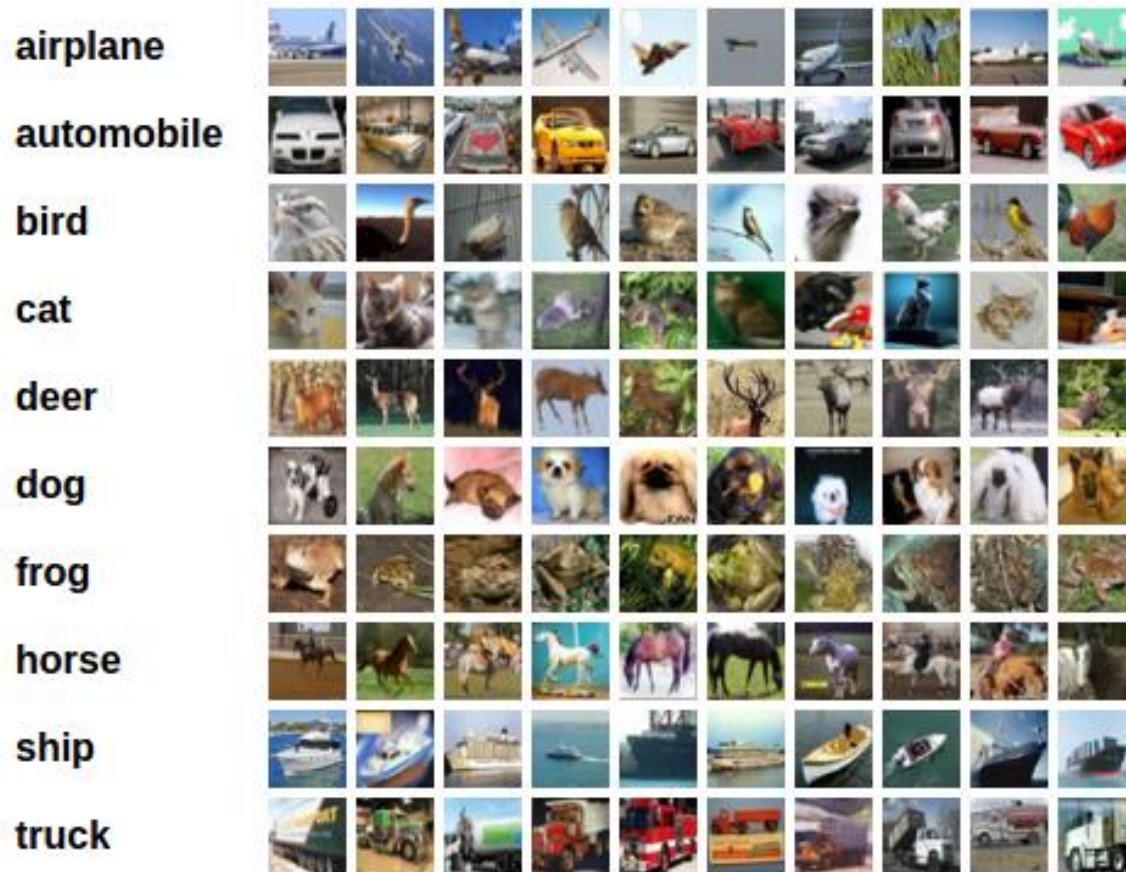
MNIST Classifier using CNN

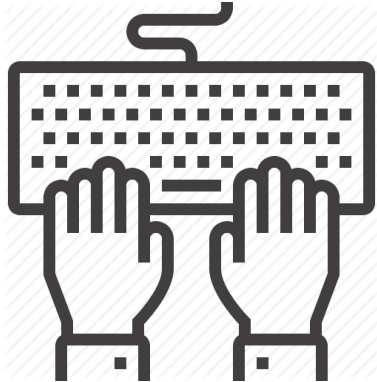
*M3.5 Convolutional Neural Network (with
inception)_MNIST.ipynb*

Projects for CNN

Image Classification using CNN

- The CIFAR-10 dataset consists of 60000 32x32 color images in 10 classes.





CIFAR10 Classifier using CNN

*M3.5 Convolutional Neural
Network_CIFAR10.ipynb*

Sentence Classification using CNN

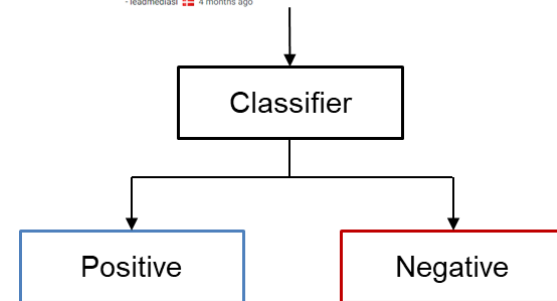
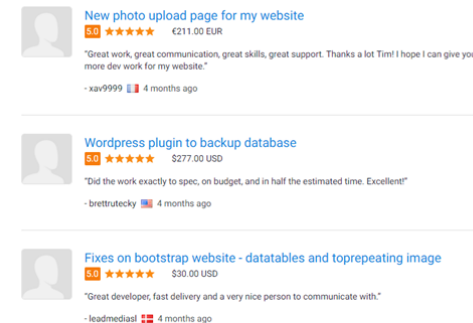
- For econominig research, classification for texts is one of the most popular applications of machine learning.

➤ Example (See Module 1.8 for more examples)

(Lee et al. 2018)

Sample Messages	Content Tags
Welcome to the unveiling of the Pro Staff RF97 that I co-designed with Wilson Tennis. Learn more at http://bit.ly/29JXLdA . #FromFederer	BRANDMENTION, PRODMENTION, PRODLLOCATION, HTTP
Coach Seve and me. Excited to be back in Brisbane! Happy we got the 1st practice of year out of the way!	SMALLTALK, EMOTION
Hello fans from Colombia! I am very happy to see you at the exo I am playing vs Tsonga on Saturday, December 15th! Buy your tickets starting September 12th on www.tuboleta.com . I hope to see you all there!	EMOTION, SMALLTALK, TARGET, PRODAVAIL, PRODLLOCATION, PRODMENTION, HTTP
The Walking Dead Season 1 DVD/Blu-ray is now available, purchase it now!!! http://blogs.amctv.com/the-walking-dead/2011/03/season-1-dvd-blu-ray.php	BRANDMENTION, PRODMENTION, PRODAVAIL, PRODLLOCATION, HTTP
Daryl makes a funny. What are some of your favorite #TheWalkingDead quotes? The highest rated quote will be turned into a graphic! #tbt	SMALLTALK, EMOTION, QUESTION, BRANDMENTION, ASKCOMMENT

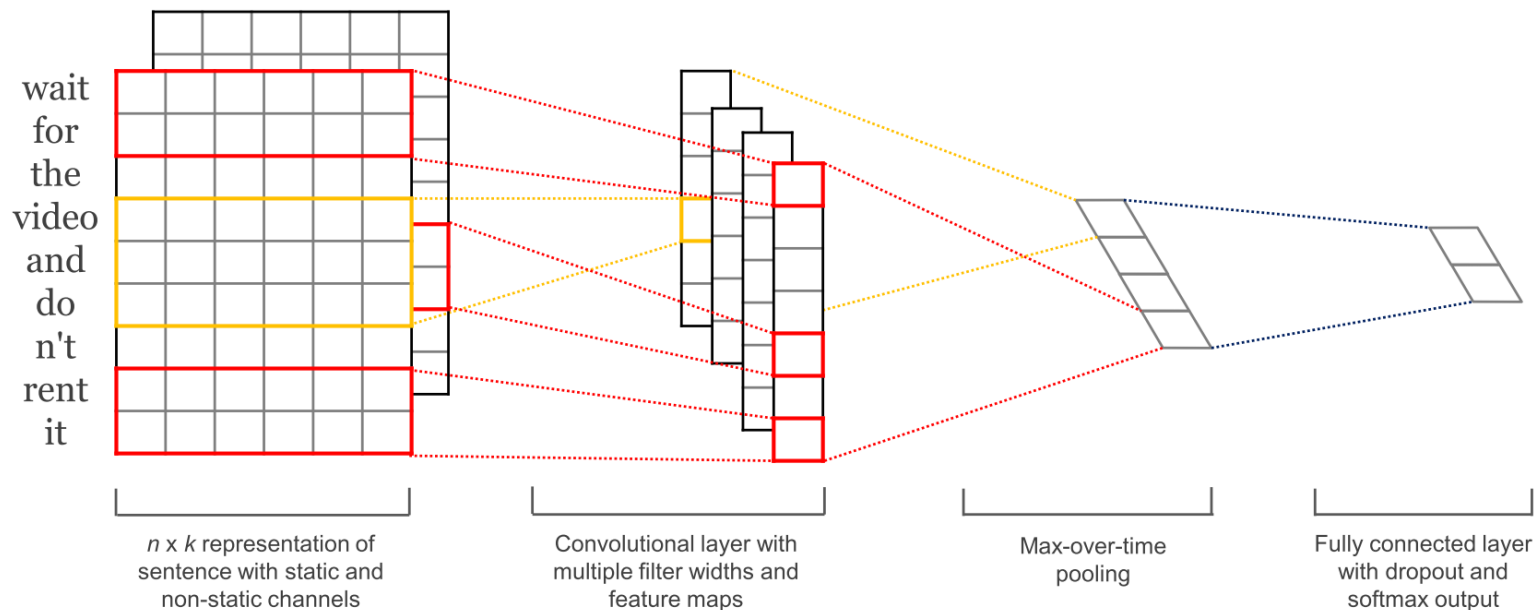
(Moreno and Terwiesch 2014)



Lee, D., Hosanagar, K. and Nair, H.S., 2018. Advertising Content and Consumer Engagement on Social Media: Evidence from Facebook. *Management Science*. forthcoming
 Moreno, A. and Terwiesch, C., 2014. Doing Business with Strangers: Reputation in Online Service Marketplaces. *Information Systems Research*, 25(4), pp.865-886.

Sentence Classification using CNN

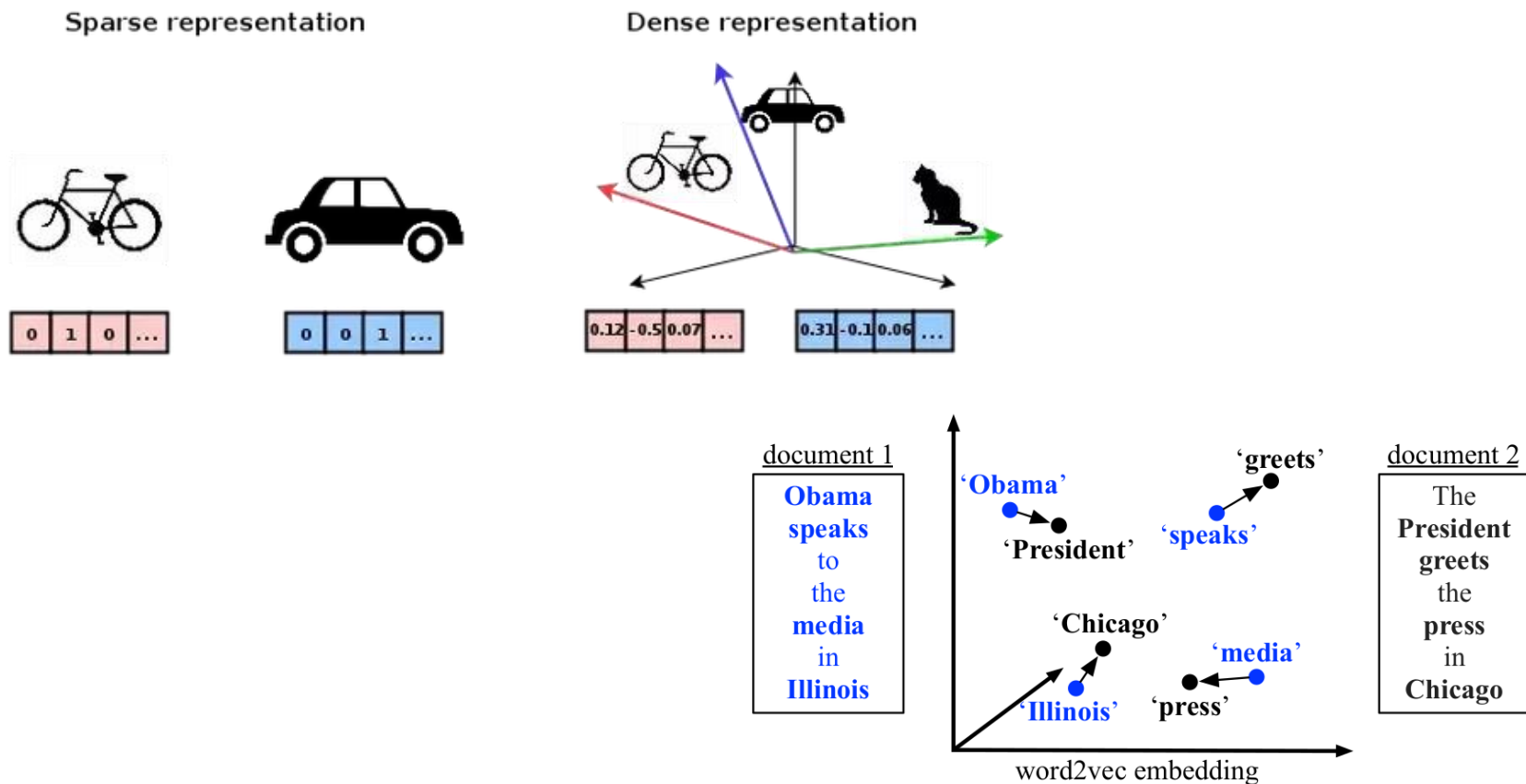
- CNN performs well in sentence classification (Kim 2014)
 - Pre-trained data is also helpful (e.g., word2vec trained on Wikipedia)

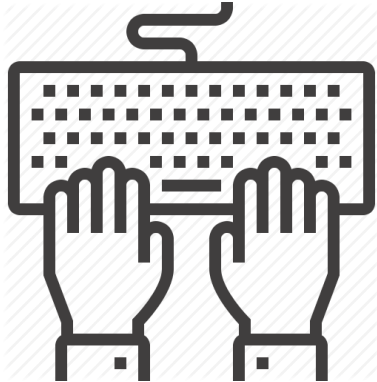


Kim, Y., 2014. Convolutional Neural Networks for Sentence Classification. In *Proceedings of the 2014 Conference on Empirical Methods in Natural Language Processing (EMNLP)*.

Sentence Classification using CNN

- Word embedding (dense representation)
 - Word vectors can represent semantic relationship between words.





Sentence Classification using CNN

*M3.5 Convolutional Neural Network_Sentence
Classification.ipynb*

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