

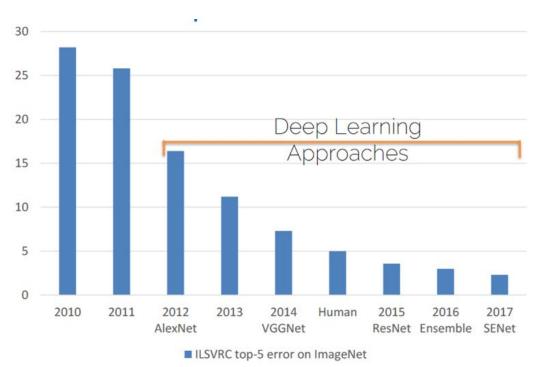
Deep Learning Software

02476 Machine Learning Operations Nicki Skafte Detlefsen



The Deep Learning Revolution







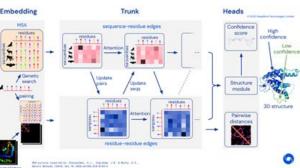
Since then



AlphaGO: Beating humans



GPT-3: Having conversations



Alphafold: Solving protein engineering



U-net: Real-time semantic segmentation

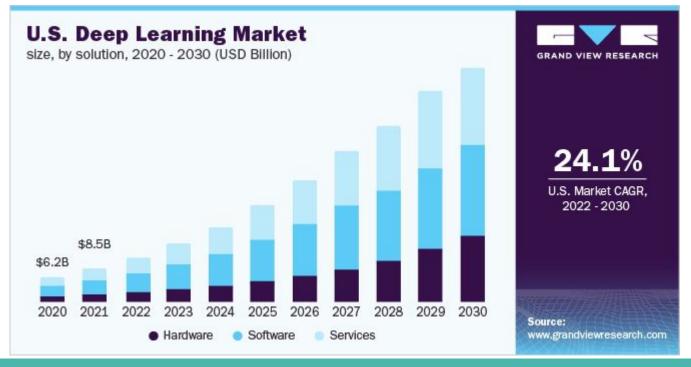


Stable diffusion: Text to image translation



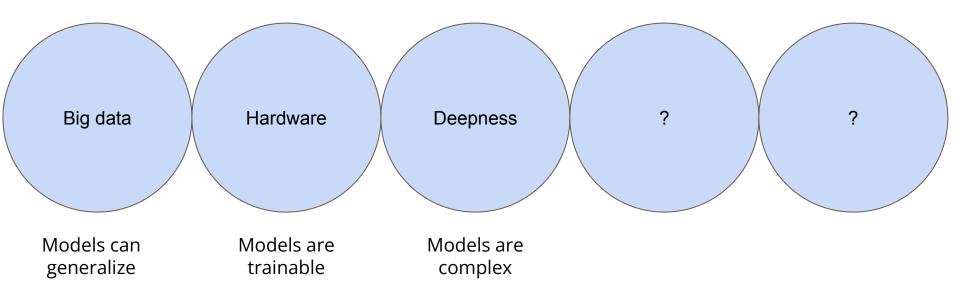
Why you should care

Usage of Deep learning in industry is increasing fast!



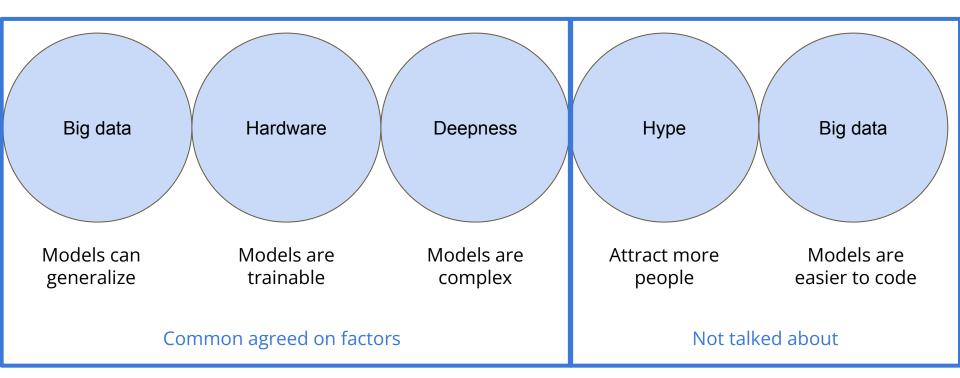


The drivers of the revolution





The drivers of the revolution





Why do we need frameworks for DL?

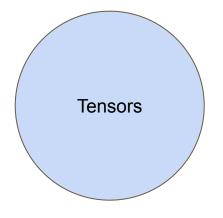
Deep learning is a lot of simple math

- 1. But we need to do it efficiently
- We need to take care of hardware acceleration (=CUDA)
- 3. We need to take care of gradient backprob
- 4. Optimizers, data interface etc. also complicates thing

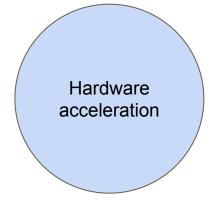
We do not really want to deal with...

```
import numpy as np
class Linear(object):
   def __init__(self, input_dim: int, num_hidden: int = 1):
        self.weight = np.random.randn(input_dim, num_hidden)
        self.bias = np.zeros(num hidden)
   def __call__(self, x):
        self.x = x
       output = x @ self.weight + self.bias
        return output
   def backward(self, gradient):
       self.weight_gradient = self.x.T @ gradient
       self.bias_gradient = gradient.sum(axis=0)
        self.x gradient = gradient @ self.weight
        return self.x_gradient
   def update(self, lr):
       self.weight = self.weight - lr * self.weight gradient
       self.bias = self.bias - lr * self.bias_gradient
if name == " main ":
   x = np.random.randn(10, 5)
   layer = Linear(5, 1)
   y = layer(x)
   grad = layer.backward(np.ones((10, 5)))
    layer.update(1e-2)
```

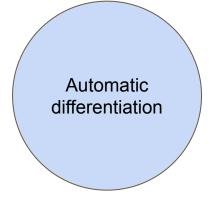




Abstraction to higher order data

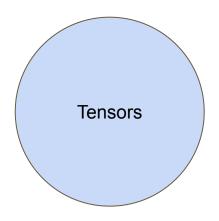


Faster computations

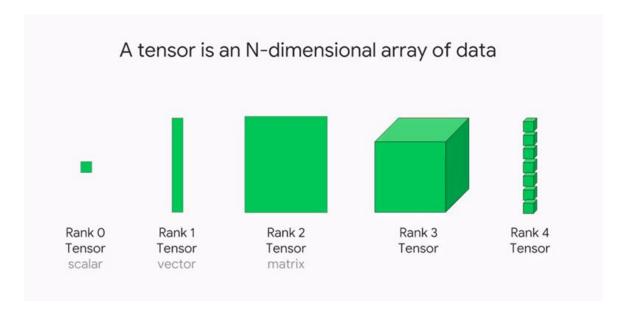


Ease of use

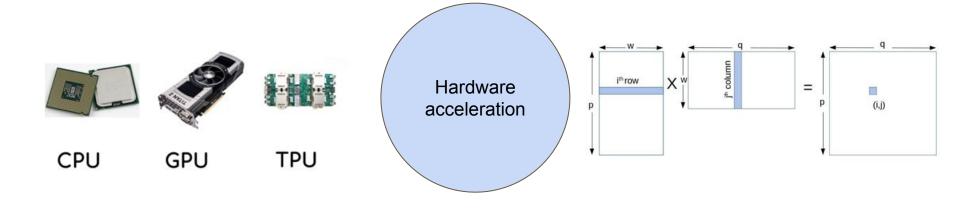




Abstraction to higher order data

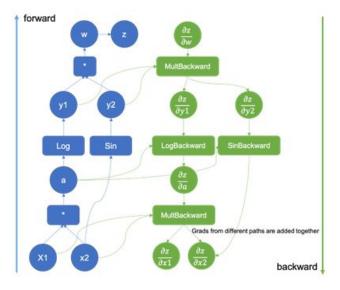






Faster computations





```
(base) C:\Users\nsde>python

Python 3.8.5 (default, Sep 3 2020, 21:29:08) [MSC v.1916 64 bit (AMD64)] :: Anaconda, Inc. on win32

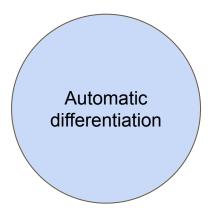
Type "help", "copyright", "credits" or "license" for more information.

>>> import torch

>>> 2*torch.ones(5, requires_grad=True)

tensor([2., 2., 2., 2.], grad_fn=<MulBackward0>)

>>>
```



Ease of use



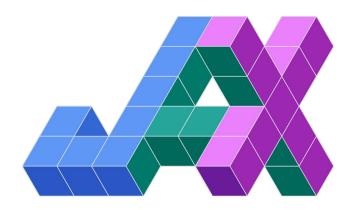
The current landscape

Pytorch / Tensorflow / Jax all supply the same

- Python interface
- Hardware acceleration
- Research and industry specific features





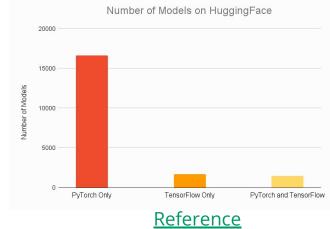




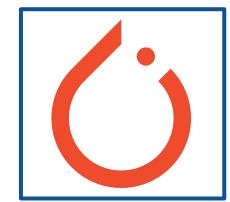
The current landscape

In this course we work with Pytorch because

- Absolutely dominant framework (#models, #papers, #competitions winners etc.)
- What we use in our research







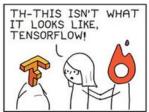


We highly recommend...

If you have the time, learn the basic of them all:)



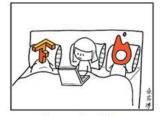












Remember, it's not a competition.



In practice, people often use high-level frameworks

Makes a lot of coding much easier.

Recommend, to only use these if you understand the underlying framework.

We get back top one of these.











Meme of the day

