

구현을 위한 딥러닝

- 고려대학교 물리학과 한승희

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1

Introduction

Self Driving Car



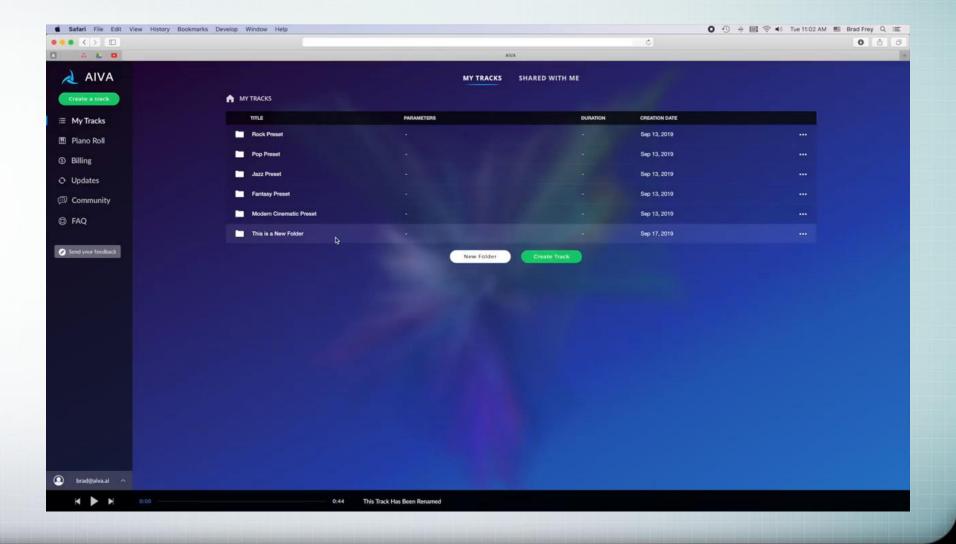
Self Driving Car



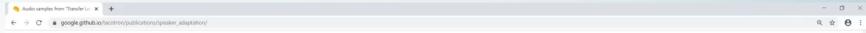
Amazon Prime Air



AIVA (AI Virtual Artist)



AI Speech (2018)



Audio samples from "Transfer Learning from Speaker Verification to Multispeaker Text-To-Speech Synthesis"

Paper: arXiv

Authors: Ye Jia *, Yu Zhang *, Ron J. Weiss *, Quan Wang, Jonathan Shen, Fei Ren, Zhifeng Chen, Patrick Nguyen, Ruoming Pang, Ignacio Lopez Moreno, Yonghui Wu. (*: equal contribution.)

Abstract: We describe a neural network-based system for text-to-speech (TTS) synthesis that is able to generate speech audio in the voice of many different speakers, including those unseen during training. Our system consists of three independently trained components:
(1) a speaker encoder network, trained on a speaker verificiation task using an independent dataset of noisy speech from thousands of speakers without transcripts, to generate a fixed-dimensional embeddending vector from seconds peech from a target speaker;
(2) a sequence-to-sequence synthesis network based on Tacotron 2, which generates a mel spectrogram from text, conditioned on the speaker embedding;
(3) an auto-regressive WaveNet-based vocoder that converts the mel spectrogram into a sequence of time domain waveform samples. We demonstrate that the proposed model is able to transfer the knowledge of speaker variability learned by the discriminatively-trained speaker encoder to the new task, and is able to synthesize natural speech from speakers that were not seen during training. We quantify the importance of training the speaker encoder on a large and diverse speaker set in order to obtain the best generalization performance. Finally, we show that randomly sampled speaker embeddings can be used to synthesize speech in the voice of novel speakers dissimilar from those used in training, indicating that the model has learned a high quality speaker representation.

Click here for more from the Tacotron team.

Speaker Adaptation for Unseen Speakers

Each column corresponds to a single speaker. The speaker name is in "Dataset SpeakerID" format. All speakers are unseen during training. The first row is the reference audio used to compute the speaker embedding. The rows below that are synthesized by our model using that speaker embedding.

These examples are sampled from the evaluation set for Table 1 and Table 2 in the paper.

VCTK p240 Reference:	VCTK p260	LibriSpeech 1320	LibriSpeech 3575	LibriSpeech 6829	LibriSpeech 8230
₹ 0:00 / 0:04	▶ 0:00 / 0:05	▶ 0:00 / 0:04	► 0:00 / 0:04 • :	► 0:00 / 0:05 — • • :	► 0.00 / 0.05 → • • • •
Synthesized: 0: Take a look at these pages for crooked creek drive.					
► 0:00 / 0:02 — • • :	▶ 0:00 / 0:02 → • • •	▶ 0:00 / 0:03 → • •	► 0:00 / 0:03 — • • • •	► 0:00 / 0:02 — • • i	▶ 0:00 / 0:03
1: There are several listings for gas station.					
▶ 0:00 / 0:02 • :	▶ 0:00 / 0:01 	▶ 0:00 / 0:02 → • • •	► 0:00 / 0:02 ● :	▶ 0:00 / 0:02 → • :	▶ 0:00 / 0:02 → • • •
2: Here's the forecast for the next four days.					
▶ 0:00 / 0:02 ● :	▶ 0:00 / 0:02 → • :	▶ 0:00 / 0:02 → • • •	▶ 0:00 / 0:02 ● :	► 0:00 / 0:02 — • • :	▶ 0:00 / 0:02
3: Here is some information about the Gospel of John.					
► 0:00 / 0:02 ● :	► 0:00 / 0:02 	► 0:00 / 0:02 ● :	► 0:00 / 0:03 — • • i	► 0:00 / 0:03 — • • :	▶ 0:00 / 0:03
4: His motives were more pragmatic and political.					
▶ 0:00 / 0:02	▶ 0:00 / 0:02 → • • :	► 0:00 / 0:03 — • i	► 0:00 / 0:02 — • • :	► 0:00 / 0:02 ——————————————————————————————————	Ola 2 and Weiss et al. 2019]
5: She had three brothers and two sisters.					
► 0:00 / 0:01 — • i	► 0:00 / 0:02 — • • i	▶ 0:00 / 0:02 → • •	► 0:00 / 0:02 — • • • •	▶ 0:00 / 0:02	▶ 0:00 / 0:02 30 d T C e

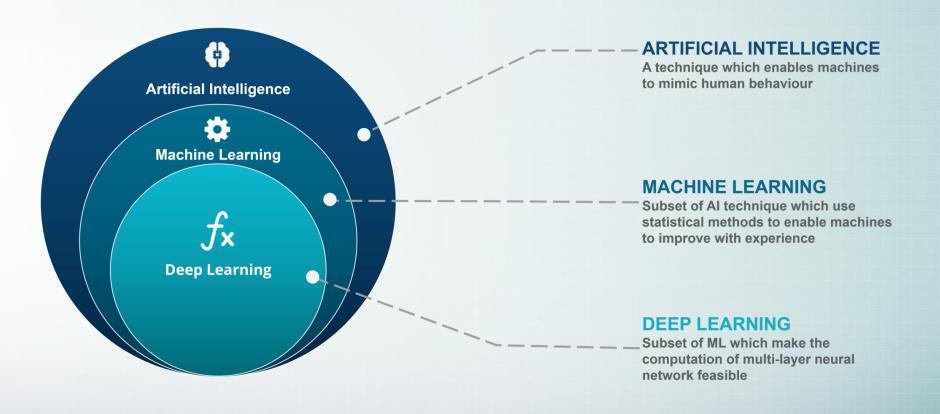
GPT - 3 (OpenAI, 2020)

```
+ Code + Text
                                                                                         Editing
≣
<>
           !pip install openai
import json
           import openai
       [ ] with open('GPT_SECRET_KEY.json') as f:
                data = json.load(f)
       [ ] openai.api key = data["API KEY"]
       [ ] from gpt import GPT
           from gpt import Example
       [ ] gpt = GPT(engine="davinci",
                     temperature=0.5,
                     max tokens=100)
```

AI Gaming

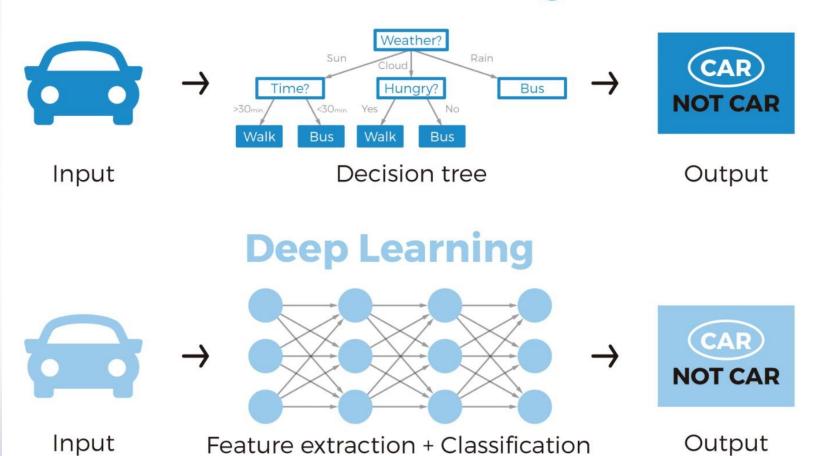
```
Gen 25 species 10 genome 7 (30%)
Fitness: 2365 Max Fitness: 2365
                               Right
          MARIO START !
```

AI vs ML vs DL



Deep Learning?

Machine Learning

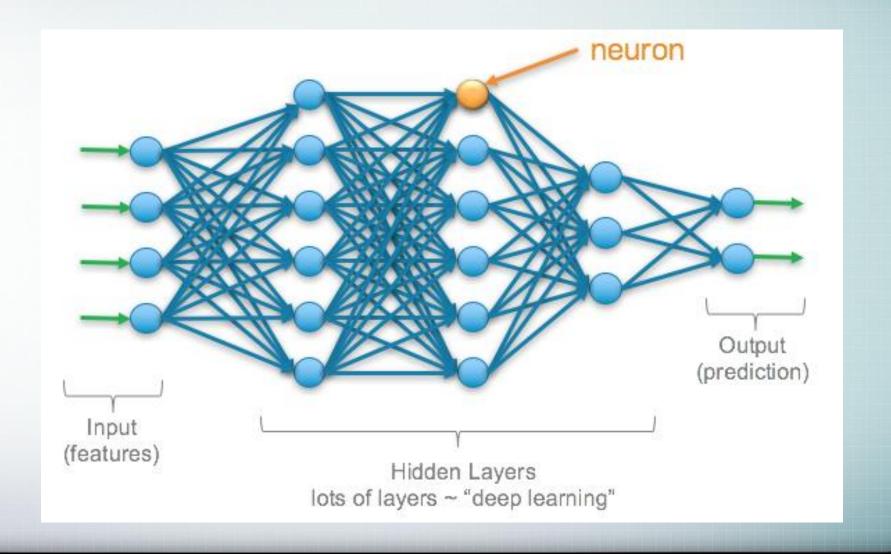


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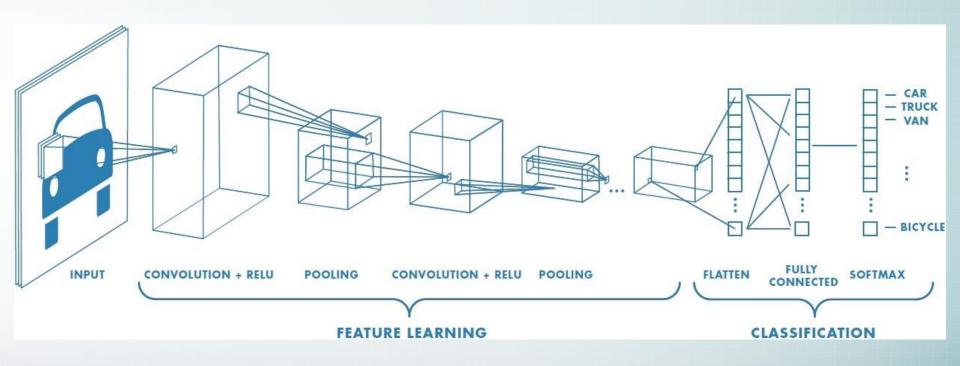
2

Deep Learning Models

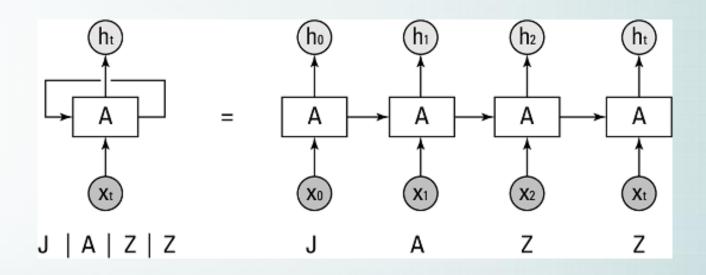
Idea of Neural Networks (NN)



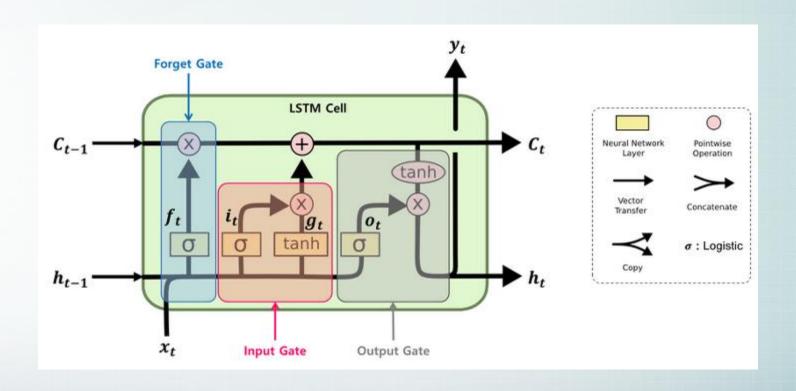
Convolutional NN (CNN)



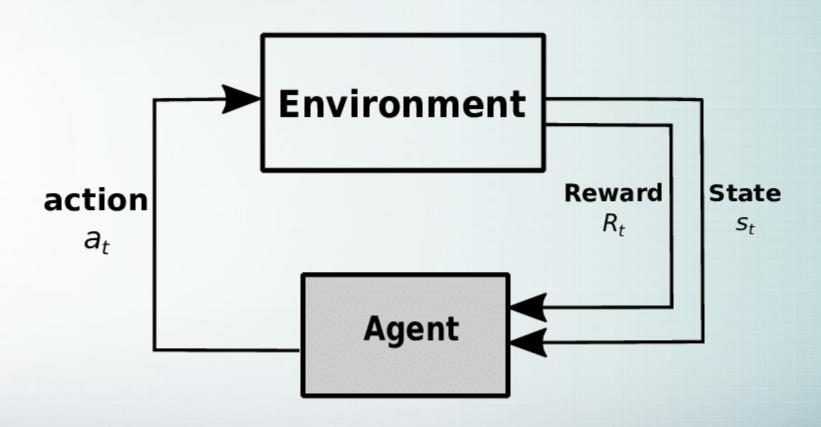
RNN



RNN - LSTM



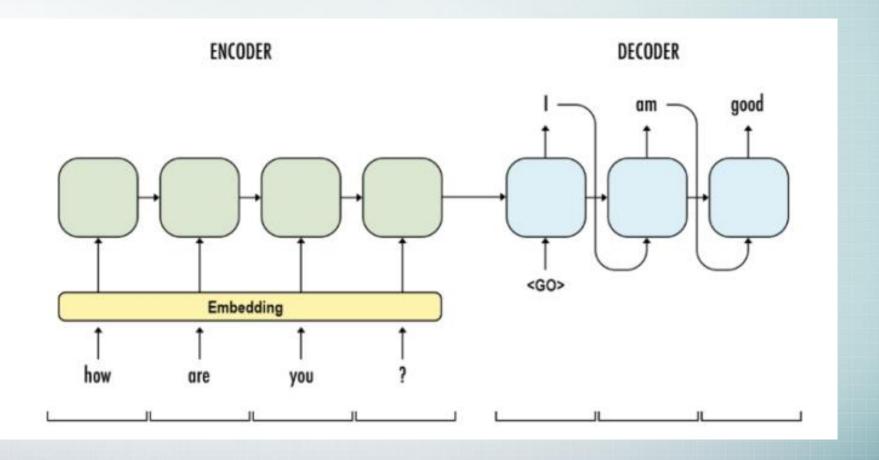
Reinforcement Learning



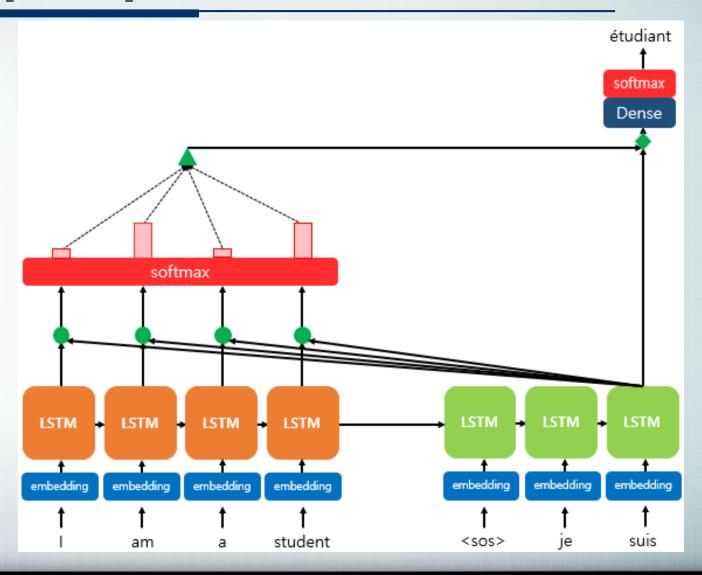
Some Important Models...

- Resnet (CNN)
- VAE (CNN + probability)
- Seq2Seq (RNN)
- Image Captioning (CNN + RNN)
- *** BERT (Transformer)**
- Image Captioning (Visual Attention)
- * XLNet (Transformer XL)
- *** GPT 3**

Seq2Seq before Attention



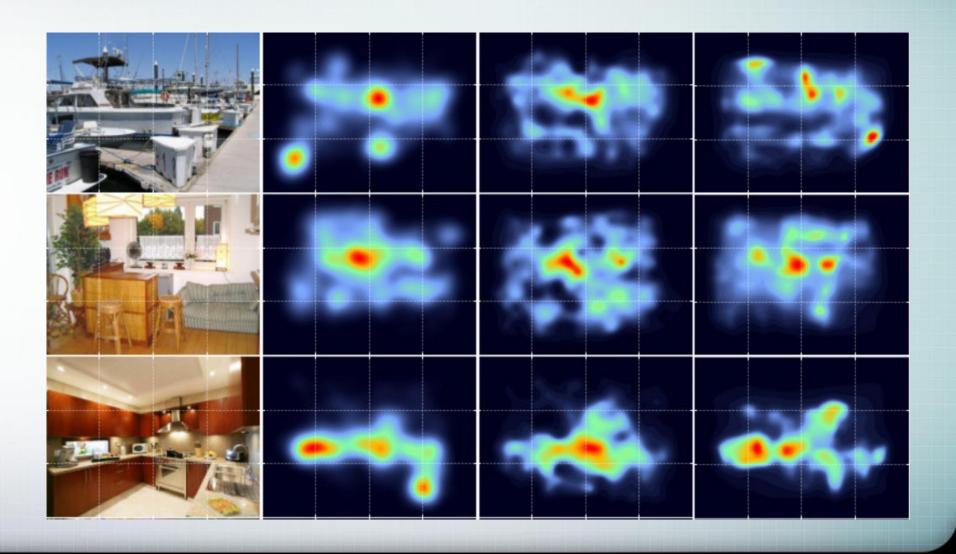
Seq2Seq with Attention



Visual Attention



Visual Attention



Transformer

Attention Is All You Need (2017) - Ashish Vaswani, Noam Shazeer, Niki Parmar, Jakob Uszkoreit, Llion Jones, Aidan N. Gomez, Lukasz Kaiser, Illia Polosukhin

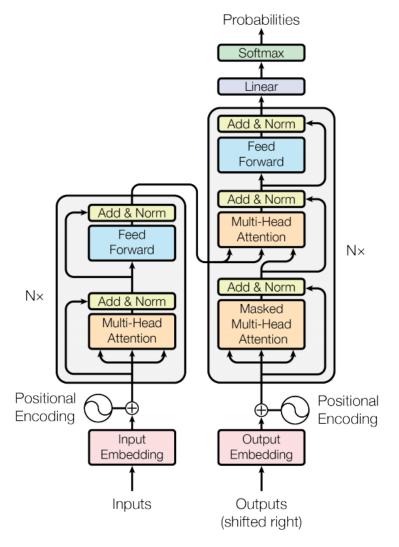


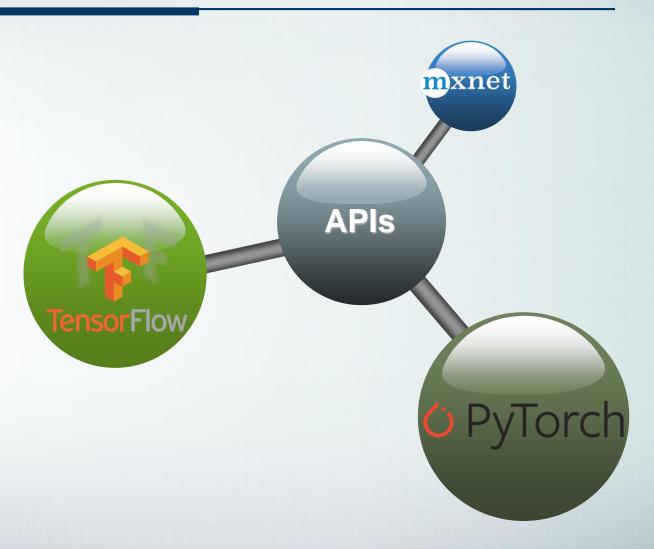
Figure 1: The Transformer - model architecture.

Contents

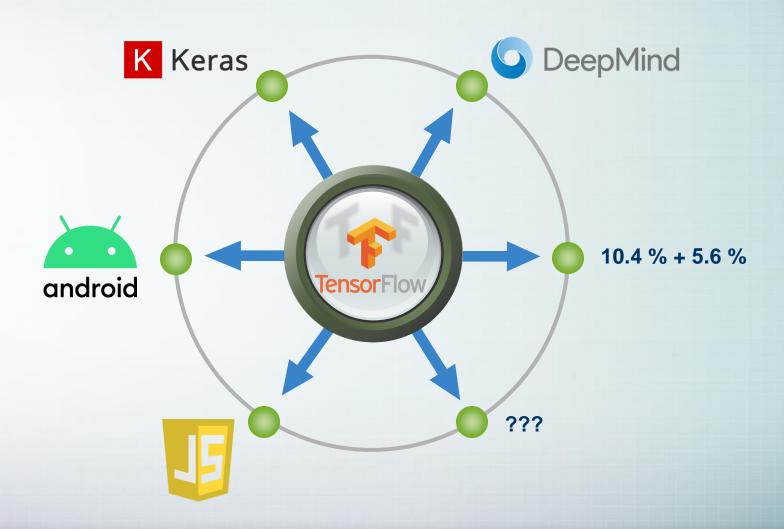
3)

Tensorflow vs PyTorch

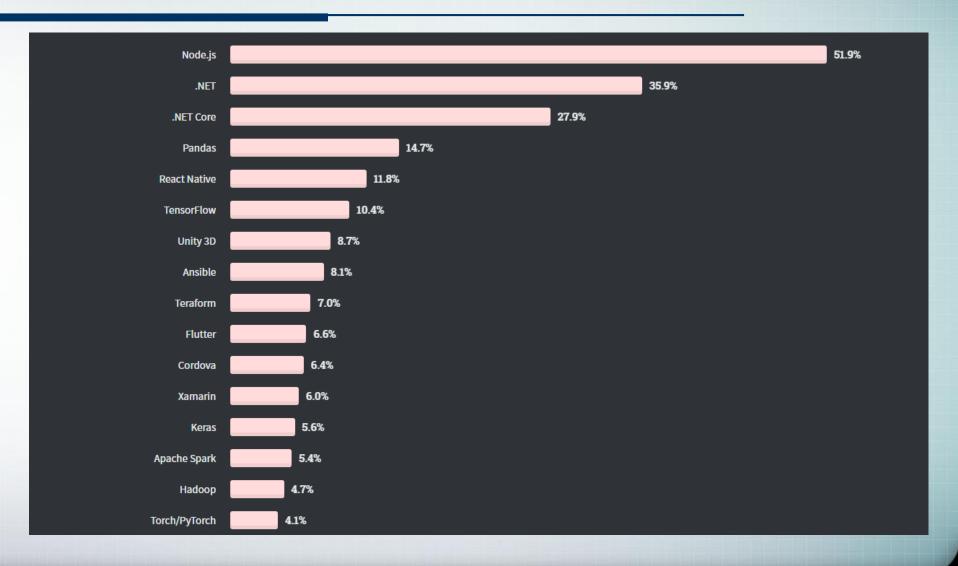
Tensorflow, PyTorch, MXNet



Why TensorFlow?



2020 Stack Overflow Developer Survey



H.W.

- 1. Run NumPy.html & Pandas.html on python (optional)
- 2. from d2l.ai (Colab recommended)
 - Run CH2.3.Linear Algebra
 - Run CH2.6.Probability
- 3. DLSeminar.pdf (optional)
 - Read Lecture 1 ~ 2 (except Lecture 2. Confusion Matrix and ...)

