# **Computational Linguistics**

Spring 2023

108.413A (002)

#### **Course Information**

- Instructor: Sangah Lee (Dept. of Linguistics, Seoul National University)
   (sanalee@snu.ac.kr)
- TA: Minji Kang (mnjkng@snu.ac.kr)
- Lecture: Tue, Thu 11:00-12:15 (bldg. 14, room 208)
- Textbooks: slides and supplementary materials provided

   + Jurafsky and Martin (2023 draft), "Speech and Language Processing"
   <a href="https://web.stanford.edu/~jurafsky/slp3/">https://web.stanford.edu/~jurafsky/slp3/</a>
- If you haven't taken the pre-requisite course [Language and Computer],
   make sure that you should be familiar with Python and language modeling:
   some traditional topics including n-grams, Naive Bayes, and Logistic Regression.

# **Objectives**

- Introduction to fundamental notions and theories on CL and NLP
  - current issues on deep learning models and the Transformer mechanism
  - large-scale language models based on them
- Development of programming and research abilities
  - understanding deep learning models and transformer mechanism
  - dealing with issues of CL and NLP
  - using Python-based tools (e.g. PyTorch)

#### Requirements

Grade Policies: Relative Grading (A-F)

Item	Attendance	Assignment	Midterm	Final	Total
Rate (%)	10	30	30	30	100

- If you have a valid reason for absence, please submit:
   the relevant documents and the Attendance Acknowledgment Request Form

   (uploaded on the eTL page)
- Midterm: paper test
- Final: project presentation
- Assignments: Python programming exercises, a taken-home quiz
- Students majoring in engineering could get separate grades.

# **Syllabus**

Week	Date	Topic
0	3/2 (Thu)	Course Introduction
1	3/7, 3/9	Basics of Text Processing and PyTorch
2	3/14, 3/16	PyTorch (Neural Networks)
3	3/21, 3/23	PyTorch (Neural Networks)
4	3/28, 3/30	Statistical Language Model (N-gram)
5	4/4, 4/6	N-gram and Entropy
6	4/11, 4/13	Text Classification
7	4/18, 4/20	Word Embedding
8	4/25, 4/27	Midterm Exam

# **Syllabus**

Week	Date	Topic	
9	5/2, 5/4	Sequence-to-Sequence Model (Encoder-Decoder)	
10	5/9, 5/11	/11 Attention	
11	5/16, 5/18	/18 Transformer	
12	5/23, 5/25 Transformer-based Pre-trained Model		
13	5/30, 6/1	HuggingFace Transformer	
14	6/8(Thu), 6/13(Tue)	Various NLP Tasks based on Transformer	
15	6/15 (Thu)	Final Project Presentations	

# **Environment Settings**

- Make sure that you have environments for Python programming:
  - Google Colab <a href="https://colab.research.google.com/">https://colab.research.google.com/</a>
  - Jupyter Notebook <a href="https://jupyter.org/">https://jupyter.org/</a>
- Assignment codes will be basically based on .ipynb.
  - .py forms are also allowed: but the paths should be correct!

#### **Supplementaries**

- Take a look over the Python libraries below!
  - NLTK (Natural Language Toolkit) <a href="https://www.nltk.org/">https://www.nltk.org/</a>
  - SpaCy <a href="https://spacy.io/">https://spacy.io/</a>
  - textacy <a href="https://textacy.readthedocs.io/en/latest/">https://textacy.readthedocs.io/en/latest/</a>
  - csv <a href="https://docs.python.org/3/library/csv.html">https://docs.python.org/3/library/csv.html</a>
  - json https://docs.python.org/3/library/json.html
- And a Python data structure "Classes" (Object Oriented Programming)
   <a href="https://docs.python.org/3/tutorial/classes.html">https://docs.python.org/3/tutorial/classes.html</a>

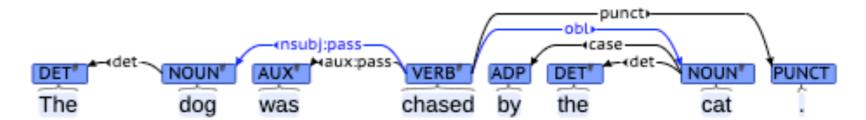
#### **Computational Linguistics**

- A subfield of linguistics and computer science
  - concerned with the interaction of human language and computers
- Includes:
  - the analysis of written texts and spoken discourse
  - the translation of text and speech from one language into another
  - the use of human languages for communication between computers and people
  - the modeling and testing of linguistic theories
- Statistical analysis of written texts and spoken discourse
  - analysis on corpus: relative frequencies or collocation of letters, sounds, morphemes, words, ...



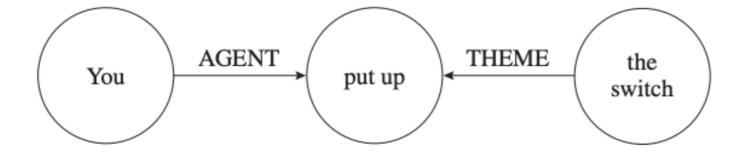
#### **Computational Linguistics**

- The interaction between language and computers in all dimensions
  - Computational Phonetics and Phonology
    - Speech Recognition, Speech Synthesis (Text-to-Speech)
  - Computational Morphology: processing of word structures
  - Computational Syntax



https://universaldependencies.org/

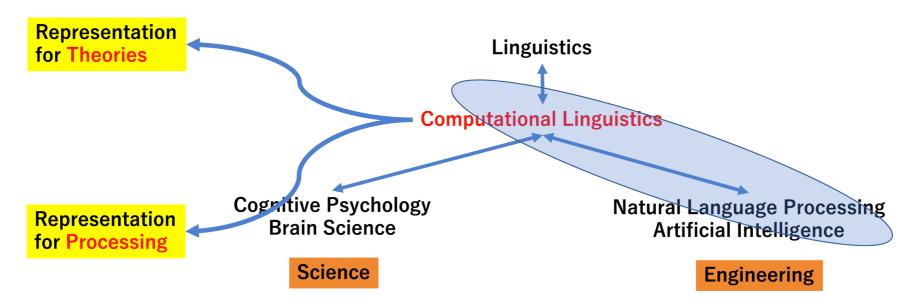
Computational Semantics: speech understanding and generation



Computational Pragmatics: sentence disambiguation, coreference resolution, …

#### **Natural Language Processing**

A schematic view of research disciplines: CL and NLP

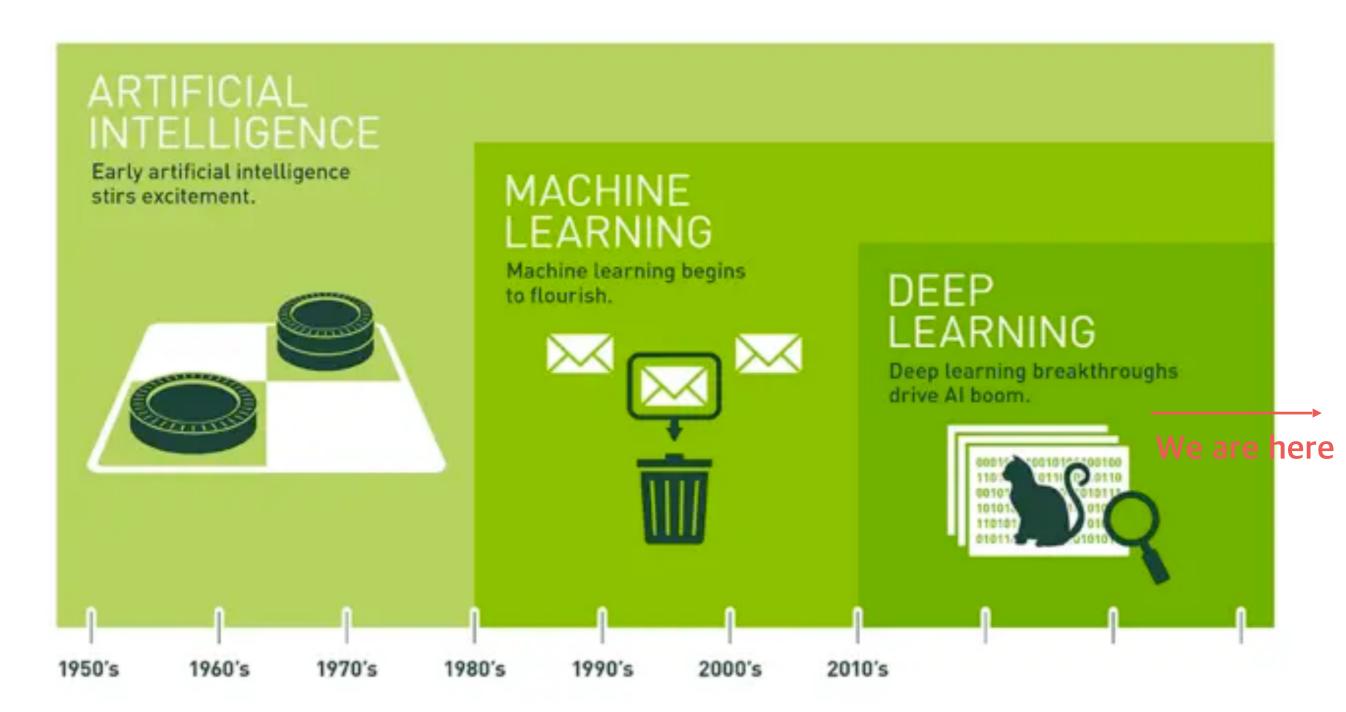


how language is processed in our minds or our brains

how computer systems should be designed to process language efficiently and effectively

NLP may be included either in CL and other fields as their subfield. (Any other fields can deal with and utilize language data.)

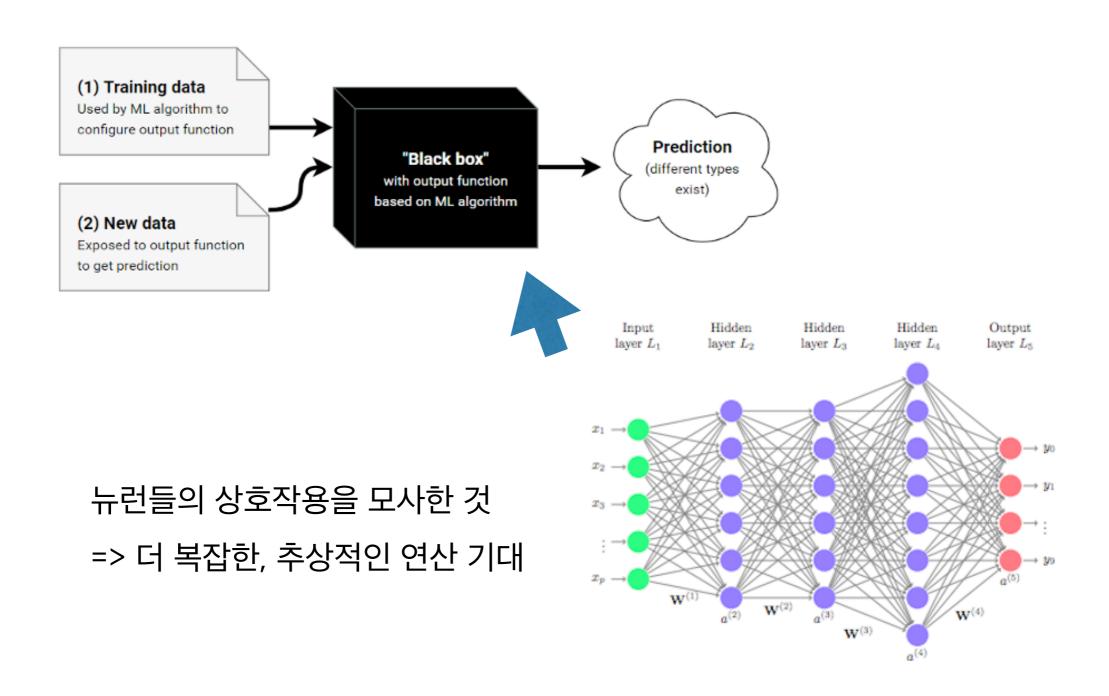
#### Deep Learning?



Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence - first machine learning, then deep learning, a subset of machine learning - have created ever larger disruptions.

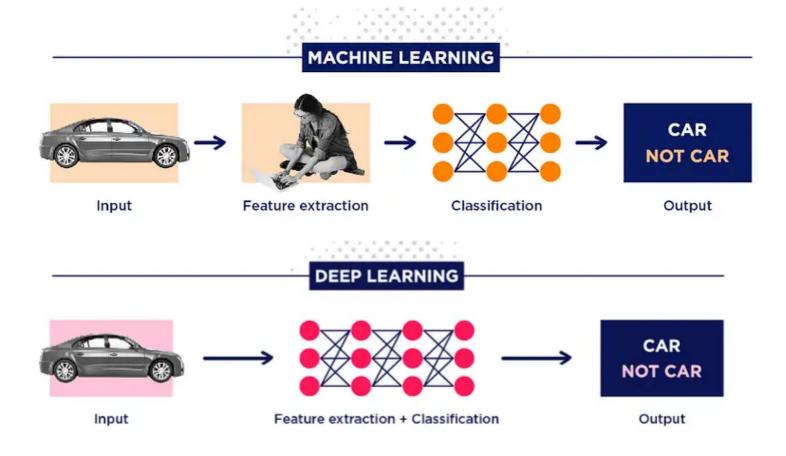
#### **Deep Learning**

● 데이터를 입력받고 그 안의 패턴과 정보를 분석하는 모델이 심층 구조로 이루어진 것



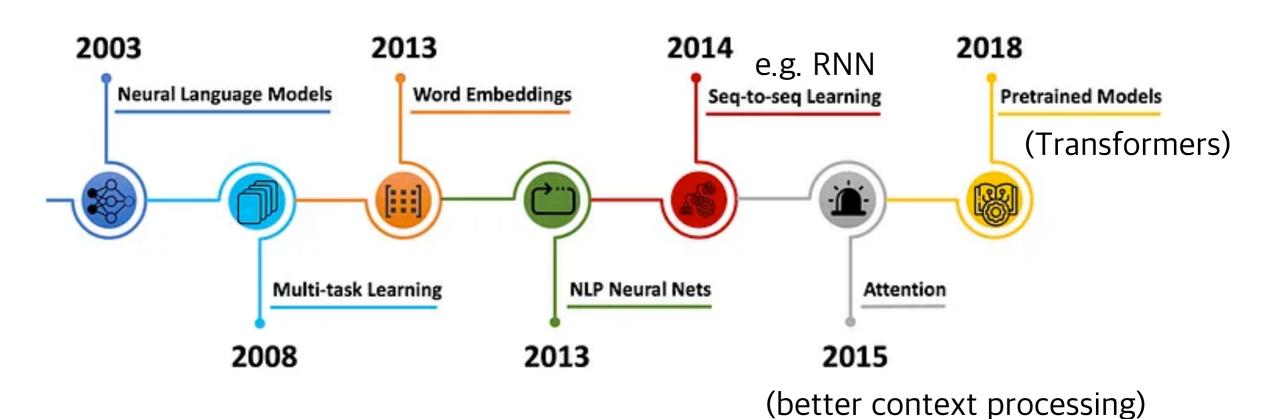
#### **Deep Learning**

- No/Less hand-crafted features required
- 기계학습 모델과 달리 사람이 직접 자질 구성을 하지 않는다. 심층학습 구조 안에서 모델이 데이터로부터 스스로 자질을 구성한다.



## Deep Learning

- In NLP:
  - Neural models automatically learn low-dimensional continuous vectors from data as task-specific features.
  - capturing semantic meanings of words, phrases, and sentences, ...
  - CNN, RNN, LSTM, ··· models for various NLP tasks



## **PyTorch**



- An open-source library for machine learning and deep learning applications
  - can use GPU accelerators -> faster!
  - 기본 자료 구조: Tensor (Numpy의 array와 유사, 호환)

return logits

● 여러 기계학습 모델, 딥러닝 모델, 학습과 검증 등에 필요한 요소 등이 구현되어 있음

```
from torch.utils.data import DataLoader

train_dataloader = DataLoader(training_data, batch_size=64, shuffle=True)

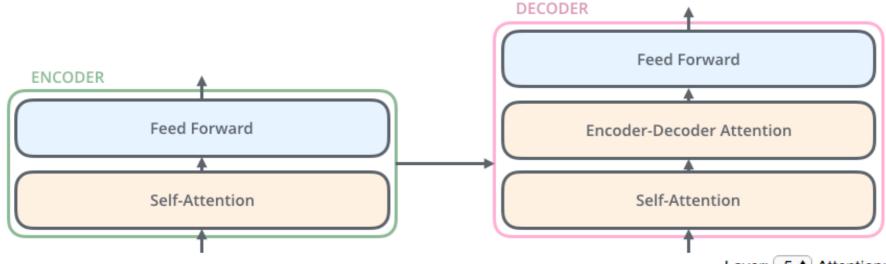
test_dataloader = DataLoader(test_data, batch_size=64, shuffle=True)

데이터 로드, 구조화
```

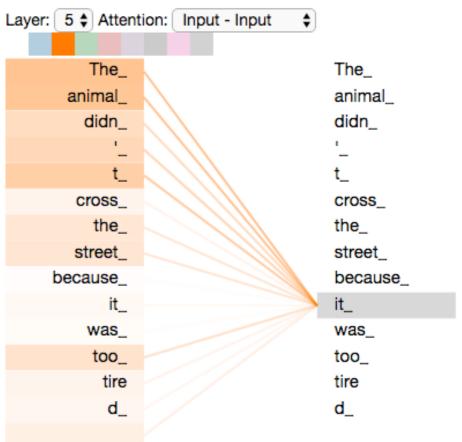
```
class NeuralNetwork(nn.Module):
                                                               class를 이용한 모델 정의
                      def init (self):
                          super().__init__()
                          self.flatten = nn.Flatten()
import torch
                          self.linear_relu_stack = nn.Sequential(
                                                                       model = NeuralNetwork()
from torch import nn
                              nn.Linear(28*28, 512),
                              nn.ReLU(),
                                                                        정의한 모델 객체 생성
                              nn.Linear(512, 512),
                              nn.ReLU(),
                              nn.Linear(512, 10),
                                                                       logits = model(X)
                          )
                                                                       모델에 input 넣고 계산
                      def forward(self, x):
                          x = self.flatten(x)
                          logits = self.linear_relu_stack(x)
```

#### **Transformers**

Using very deep neural models and attention



- Effectively modeling semantic meanings of language
  - e.g. Focusing on relationship between words based on semantic meaning and contexts

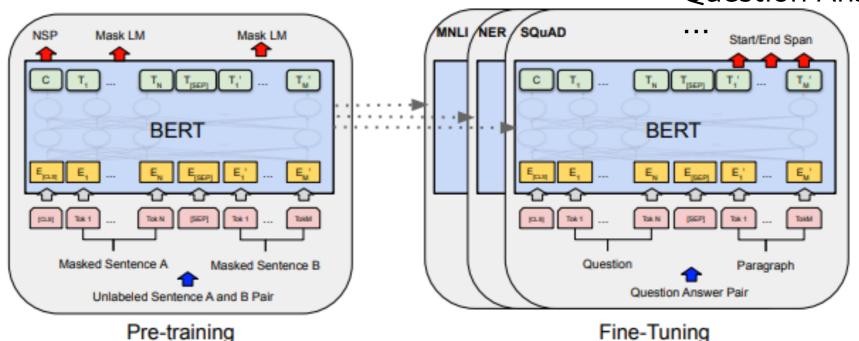


#### **Pre-Trained Models**

- Such deep and complex calculations are performed and freezed for future use
   ===> Pre-Trained (Language) Models: PTM, PLM (사전학습모델)
  - Pre-trained models: BERT, GPT, ···
  - Usually provided through HuggingFace

tasks:

Natural Language Inference Named Entity Recognition Question Answering



import and load a given BERT model

let the model perform specific task

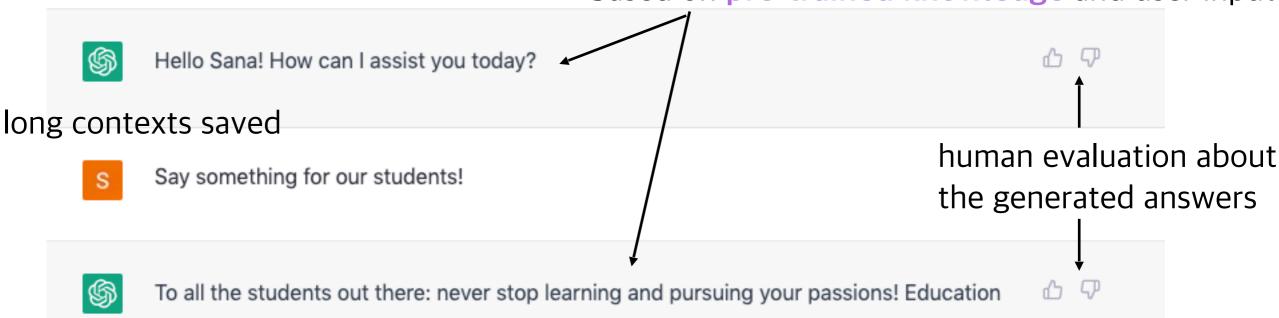
#### And you will like this

ChatGPT: one of the PTMs
 especially working for dialogues

# ChatGPT: Optimizing Language Models for Dialogue

https://openai.com/blog/chatgpt/

Hi ChatGPT, I am Sana. The model's decoders generate sentences (sequence of words) based on pre-trained knowledge and user inputs

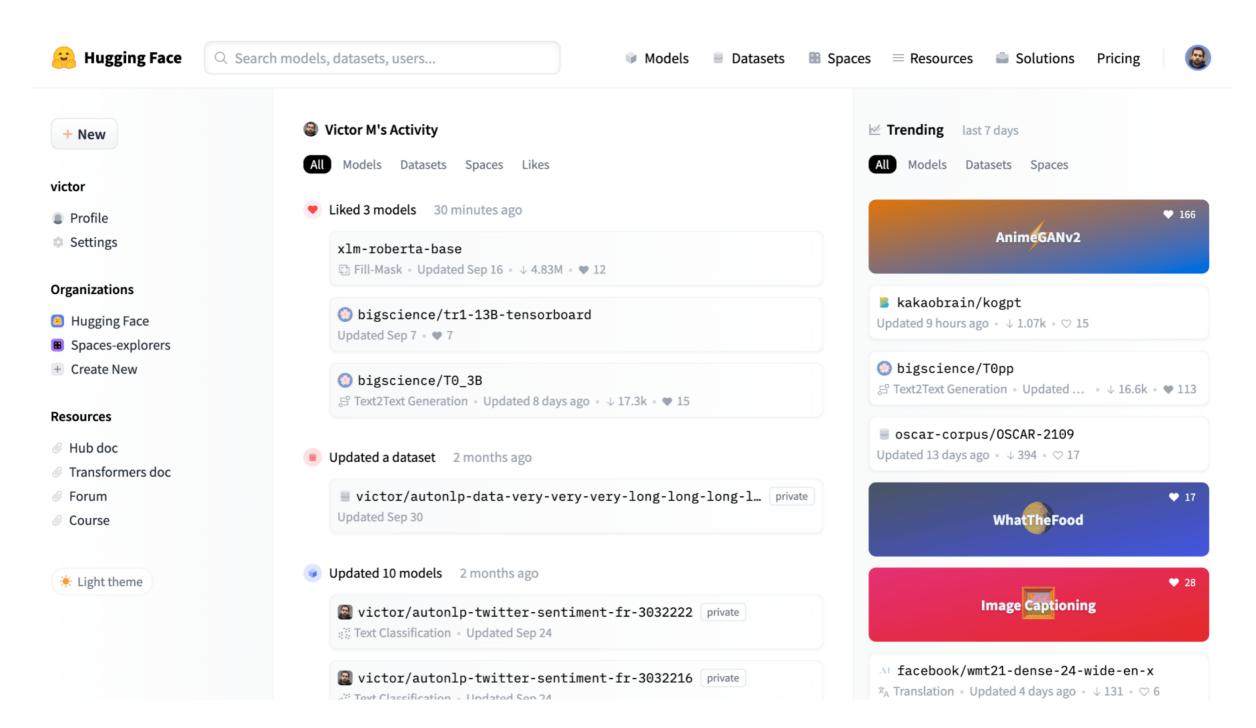


To all the students out there: never stop learning and pursuing your passions! Education opens doors to endless possibilities and empowers you to make a positive impact in the world. Stay curious, ask questions, and never be afraid to make mistakes. Remember, they are opportunities to grow and learn. Good luck on your academic journey!

# HuggingFace



A platform and a library providing various models, datasets and codes



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A platform and a library providing various models, datasets and codes

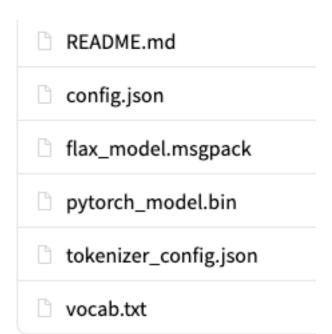


#### KoRean based Bert pre-trained (KR-BERT)

This is a release of Korean-specific, small-scale BERT models with comparable or better performances developed by Computational Linguistics Lab at Seoul National University, referenced in <a href="https://krean-specific.com/KR-BERT: A">KR-BERT: A</a>
<a href="https://krean-specific.com/KR-BERT: A">Small-Scale Korean-Specific Language Model</a>.

#### Vocab, Parameters and Data

KR-BERT sub- character	KR-BERT character	KoBERT (SKT)	KorBERT (ETRI)	Mulitlingual BERT (Google)	
12,367	16,424	8,002	30,797	119,547	vocab size
96,145,233	99,265,066	92,186,880	109,973,391	167,356,416	parameter size



# HuggingFace



[0.4552, -1.0330, -0.0385, ..., 0.1119, -0.7874, -0.1191]]]

1.7813e-01, 2.4731e-01, 1.4999e-01, -3.3565e-01, -4.1460e-01, -1.6348e-01, -6.1204e-01, 3.5326e-01, -4.8477e-01, 1.6353e-01, -7.5884e-02, -9.3399e-02, -4.9610e-02, 7.0147e-02, 8.9095e-01, -3.7699e-02, 3.6045e-01, -7.8517e-01, -3.9698e-01, 1.5543e-01, 4.0353e-01, -1.2426e-01, 1.9852e-01, 2.1575e-01, -8.6640e-02, 4.6102e-02, 3.1457e-01, 8.0098e-04, -2.7198e-01, 2.2101e-01, -3.1642e-01, 1.1386e-03, 1.4578e-01, -8.6948e-02, 4.3643e-02,

You can import and use the pre-trained models!

```
from transformers import BertConfig, BertModel, BertForPreTraining, BertTokenizer
     tokenizer = BertTokenizer.from pretrained('snunlp/KR-BERT-char16424', do lower case=False)
     sentence = '내가 만든 쿠키 너를 위해 구웠지'
     tkns = tokenizer.tokenize(sentence)
                                          Sentence Tokenization
     print(tkns)
     ['내가', '만든', '쿠', '##키', '너', '##를', '위해', '구', '##웠', '##지']
     model = BertModel.from pretrained('snunlp/KR-BERT-char16424')
      inputs = tokenizer(sentence, return tensors="pt")
      output = model(**inputs)
                                               Sentence Encoding
      print(output)
BaseModelOutputWithPoolingAndCrossAttentions(last_hidden_state=tensor([[[ 0.3934, -0.2802, -0.1499, ..., -0.0381, -0.4774,
        [0.6975, 0.1270, -0.7793, ..., 1.0838, -0.1161, -0.8736],
        [0.7390, 0.1325, -0.5397, ..., 0.5206, 0.2594, -0.8449],
        [-0.2334, 0.1795, -0.6226, ..., 0.2216, -0.3278, 0.3870],
        [-0.3575, -0.1823, -0.7299, \ldots, 0.0212, 0.0282, -0.7853],
```

grad fn=<NativeLayerNormBackward0>), pooler output=tensor([[ 2.6916e-01, -2.0484e-01, 1.6539e-01, 2.5938e-01, 2.9

#### For Next Week

Article:

https://medium.com/@ageitgey/natural-language-processing-is-fun-9a0bff37854e

Library: try PyTorch in your environment! https://pytorch.org/