

28.11.24

# Machine Learning With TensorFlow

## CNNS IN TENSORFLOW PART II

- **QUIZ**
- **OPEN DISCUSSION**
- **ASSIGNMENTS**
- **DATASET CHARACTERISTICS**

# QUIZ



<https://forms.office.com/e/0G9MpSAnid>

# OPEN DISCUSSION I

- **What is the relationship or difference between loss and accuracy?**
- **What is the relationship or difference between a batch and an epoch and what is the effect of choosing a smaller or larger batch size?**

# OPEN DISCUSSION II

- **Would you fine-tune all layers of a pre-trained model or just the head added specifically for your task?**
- **In the notebook on „Cats vs. Dogs“ the training accuracy is always lower than the validation accuracy. What are possible reasons?**

# **ASSIGNMENTS (WEEK 3 & 4)**

**ASSIGNMENTS NEXT WEEK?**

# **DATASET CHARACTERISTICS**



# **DATASET CHARACTERISTICS**

- **Write down key points on the most important aspects of how your data was collected**
- **What are potential biases?**
- **Are there outliers in the dataset?**
- **For classification tasks: Are the classes balanced?**
- **Are there potential data augmentation approaches you can use?**

# ERSTE DATENINSPEKTION

- **df.head(), df.tail():**  
**Zeigt die ersten/letzten Zeilen.**
- **df.sample():**  
**Zeigt eine zufällige Auswahl von Zeilen.**
- **df.shape:**  
**Gibt die Dimension des DataFrames zurück.**
- **df.info():**  
**Zusammenfassung des DataFrames, inkl. Datentypen.**

# DESKRIPTIVE STATISTIKEN

- **df.describe():**  
**Statistische Zusammenfassung der numerischen Spalten.**
- **df.isnull():**  
**Überprüft auf NaN-Werte.**

# VISUALISIERUNGEN

- **Scatterplots**
- **Balkendiagramme**
- **Histogramme**

# PROJECTS MILESTONES

- (1) Project pitches**
- (2) Form groups**
- (3) Literature review**
- (4) Dataset characteristics**
- (5) Baseline model**
- (6) Model evaluation**
- (7) Final model optimizations**
- (8) Project presentations**

# NEXT WEEK: BASELINE MODEL

- **How good is your final model actually? Metrics like RMSE or accuracy are difficult to interpret by themselves.**
- **A baseline model is a simple model that provides a context for the performance of your metric.**
  - **In regression problems, this might be a linear model.**
  - **In time series problems, this might be a simple forecast of the last known value.**
- **It also requires you to think about metrics, data processing, train-test splits, which you need for the complex model as well.**

# **RESOURCES FOR DATASETS AND MODELS**

## Hub

Overview

Guide

Tutorials

API

Models ↗

# TensorFlow Hub is a repository of trained machine learning models.

TensorFlow Hub is a repository of trained machine learning models ready for fine-tuning and deployable anywhere. Reuse trained models like BERT and Faster R-CNN with just a few lines of code.

[See the guide](#)

Learn about how to use TensorFlow Hub and how it works.

[See tutorials](#)

Tutorials show you end-to-end examples using TensorFlow Hub.

[See models](#)

Find trained TF, TFLite, and TF.js models for your use case.

```
!pip install --upgrade tensorflow_hub
```

```
import tensorflow_hub as hub
```

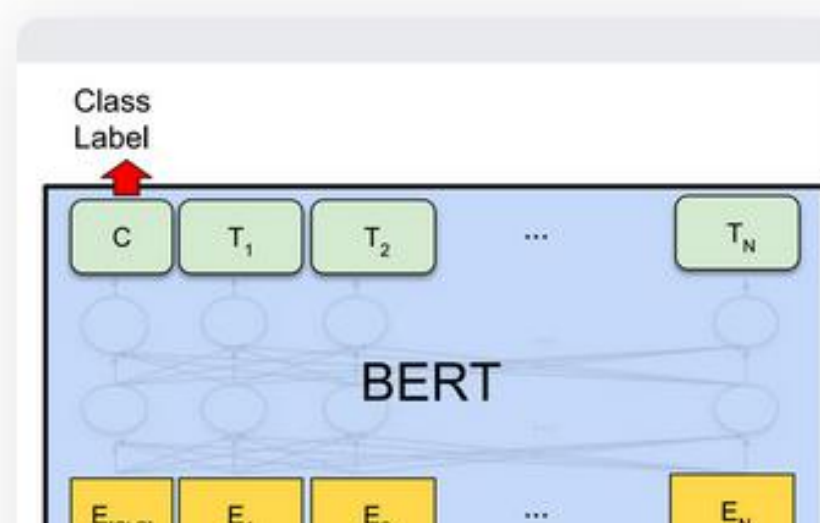
```
model = hub.KerasLayer("https://tfhub.dev/google/nnlm-en-dim128/2")  
embeddings = model(["The rain in Spain.", "falls",  
                    "mainly", "In the plain!"])
```

```
print(embeddings.shape)  #(4,128)
```



## Models

Find trained models from the TensorFlow community on [tfhub.dev](https://tfhub.dev)



Food V1.1







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SAMURAI: Adapting Segment Anything Model for Zero-Shot Visual Tracking with Motion-Aware Memory

yangchris11/samurai • PyTorch • 18 Nov 2024

The Segment Anything Model 2 (SAM 2) has demonstrated strong performance in object segmentation tasks but faces challenges in visual object tracking, particularly when managing crowded scenes with fast-moving or self-occluding objects.

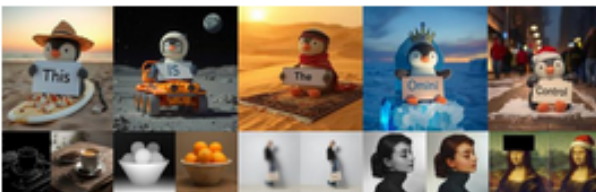
Ranked #1 on [Visual Object Tracking on GOT-10k](#)

- Visual Object Tracking
- Visual Tracking

★ 4,895

14.04 stars / hour

- Paper
- Code



OminiControl: Minimal and Universal Control for Diffusion Transformer

Yuanshi9815/OminiControl • PyTorch • 22 Nov 2024

In this paper, we introduce OminiControl, a highly versatile and parameter-efficient framework that integrates image conditions into pre-trained Diffusion Transformer (DiT) models.

★ 506

3.97 stars / hour

- Paper
- Code



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- Text-to-Video Visual Question Answering
- Document Question Answering Graph Machine Learning

Computer Vision

- Depth Estimation Image Classification
- Object Detection Image Segmentation
- Image-to-Image Unconditional Image Generation
- Video Classification Zero-Shot Image Classification

Natural Language Processing

- Text Classification Token Classification
- Table Question Answering Question Answering
- Zero-Shot Classification Translation
- Summarization Conversational
- Text Generation Text2Text Generation
- Sentence Similarity

Audio

- Text-to-Speech Automatic Speech Recognition
- Audio-to-Audio Audio Classification
- Voice Activity Detection

Tabular

- Tabular Classification Tabular Regression

Reinforcement Learning

- Reinforcement Learning Robotics

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meta-llama/Llama-2-70b  
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cerspense/zeroscope\_v2\_XL  
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# ADDITIONAL DISCUSSION QUESTIONS

**Explain how the number of parameters are set together (Convolutions are defined to use 3x3 filters and Poolings to use 2x2)**

Layer (type)	Output Shape	Param #
conv2d_12 (Conv2D)	(None, 26, 26, 64)	640
max_pooling2d_12 (MaxPooling)	(None, 13, 13, 64)	0
conv2d_13 (Conv2D)	(None, 11, 11, 64)	36928
max_pooling2d_13 (MaxPooling)	(None, 5, 5, 64)	0
flatten_5 (Flatten)	(None, 1600)	0
dense_10 (Dense)	(None, 128)	204928
dense_11 (Dense)	(None, 10)	1290

# TASKS UNTIL NEXT WEEK

- **Complete of week 1 and week 2 of the course Natural Language Processing in TensorFlow**
- **Complete the exercise provided in the course handbook.**
- **Formulate a baseline model**