

# Preposition Sense Disambiguation

*TTLab - Text2Scene Praktikum*

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# Aufgabe

# Motivation

Preposition-Sinn wichtig für Verständnis:

Auf *der Wache* bedeutet eigentlich, dass man **in** dem Gebäude ist

Auf *dem Tisch* bedeutet, dass etwas wirklich **auf** dem Tisch ist

Zur Visualisierung braucht man diese Informationen

# Nice to Know

## SemEval Benchmark:

- **SemEval-Winner 2007** - Max-Entropy-Ansatz: Accuracy bis 75 %
- **Litkowski 2013** - Lemmatizer und Dependency-Parser: Accuracy: 86 %
- **Semi-Supervised(2016)** - Ausnutzen von mehreren Sprachen: Accuracy bis 80 %
- **Hongyu Gong(2018)** - Geographische Context-Vektoren: Accuracy bis 80 %

# Unsere Aufgabe

- Datenbeschaffung und -bereinigung
- Trainieren von 2 Taggern (hyperparameter-optimiert)
- Einbinden in den Text-Imager
- Semi-Supervised Ansatz Wenn möglich einbinden

## Datenbeschaffung und -bereinigung (Tim und Dirk)

# Daten

- SemEval 2007 oft als Benchmark Datensatz
- schwer zu bekommen online
- 34 englische Präpositionen mit insgesamt 224 verschiedene Sinnen
- Je preposition 2 xml (Trainingsdata und label-defintionen)



# Datenbeispiel

## Trainingsdata:

```

<instance id="with.p.fn.340887" docsrc="FN">
  <answer instance="with.p.fn.340887" senseid="7(5)"/>
  <context>
    Scott threw himself <head>with</head> enthusiasm into this exacting assignment .
  </context>
</instance>

<instance id="with.p.fn.342218" docsrc="FN">
  <answer instance="with.p.fn.342218" senseid="11(7b) 7(5)"/>
  <context>
    He took another sniff , wrinkling his nose <head>with</head> distaste .
  </context>
</instance>

```

## Definitionen:

```

<sense id="5">
  <definition>because of / due to (the physical/mental presence of) (e.g., boiling with anger, shining with dew)</definition>
  <majorcluster>CAUSE</majorcluster>
  <pprojmap type="equivalent" targetid="11(7b)"/>
  <notes></notes>
</sense>
<sense id="6">
  <definition>indicating the manner or circumstances (but not cause or motivation) of something (e.g., fix with precision)</definition>
  <majorcluster>MANNER</majorcluster>
  <pprojmap type="equivalent" targetid="7(5)"/>
  <notes></notes>
</sense>

```

# Aufbereitete Daten

sentence_id	sentence	label_id	definition
0	The USSR 's fifteen union republics , united <head-on</head> a supposedly voluntary basis , formed an ' integral , federal , multinati	22	having (the thing mentioned) as criteria used in judgment or evaluation or as a/the just
1	Taking usage rate as a variable essentially means segmenting <head-on</head> the basis of volume purchased .	22	having (the thing mentioned) as criteria used in judgment or evaluation or as a/the just
2	She pinched bruises <head-on</head> her daughter 's inner arm , and had poured hot tea on both daughters .	1	indicating a surface that , due to contact with , serves as the cause or means of causing
3	Gingerly I squeezed a bit <head-on</head> my fingertip .	1	indicating a surface that , due to contact with , serves as the cause or means of causing
4	Tom squeezed <head-on</head> the reins and they came to a halt .	13	physically in contact with and supported by (a surface) (e.g., the book on the table)
5	I pointed my piece at Johnny and squeezed <head-on</head> the trigger .	13	physically in contact with and supported by (a surface) (e.g., the book on the table)
6	Madeleine and <u>Victorine</u> stood behind his wheelchair , like nurses , while the two girls huddled <head-on</head> the <u>windowseat</u> .	13	physically in contact with and supported by (a surface) (e.g., the book on the table)
7	She would kneel <head-on</head> the floor behind the sofa and pull out the tall books of art reproductions .	13	physically in contact with and supported by (a surface) (e.g., the book on the table)
8	She knelt <head-on</head> the cold stone floor and carefully placed some coals on the dying embers in the grate .	13	physically in contact with and supported by (a surface) (e.g., the book on the table)
9	Sung was kneeling <head-on</head> the top of the dyke , staring across at the House as the dawn broke .	13	physically in contact with and supported by (a surface) (e.g., the book on the table)
10	The premises are freehold , " he continued , not responding to my comment , but at least no longer leaning <head-on</head> the co	13	physically in contact with and supported by (a surface) (e.g., the book on the table)

# Huggingface-Bert (Dirk)

# Huggingface-Bert

Huggingface library:

- Leichte Umsetzung von State-of-the-Art Modellen
- Unterstützt bert

Bert:

- Released: November 2019
- Sehr(!) gut vortrainiert
- Knackt gleich in mehreren Disziplinen die State-of-the-Art (GLUE, SQuAD, SWAG)

# Huggingface-Bert-Trainer

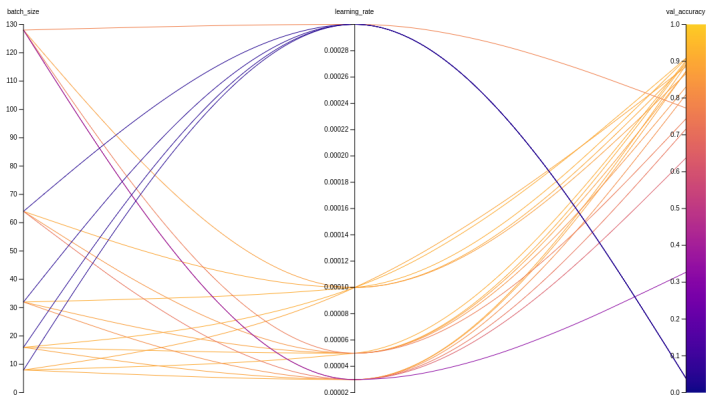
- 1 Daten einlesen (90:10 train:val Split)
- 2 Daten tokenisieren (pretrained tokenizer von hf für Bert):  
Sätze werden zu **Input-Ids** und **Attention-Masks**
- 3 bert Modell initialisieren als **Klassifizierungsmodell** (mit 224 verschiedenen Outputs)
- 4 Trainingsschleife

# Huggingface-Bert-Hyperparameter-Optimierung

Google empfiehlt

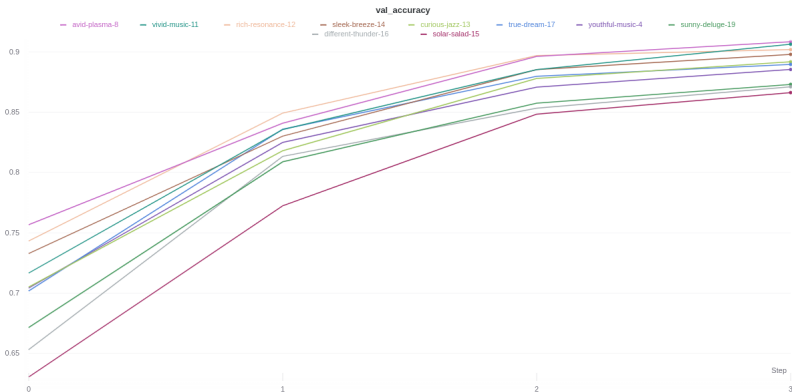
- Epochen: 4
- Optimizer: Adam
- Learning-Rate aus [3e-4, 1e-4, 5e-5, 3e-5]
- Batch-Size aus [8, 16, 32, 64, 128]
- Rest: schon vorgegeben

# Huggingface-Bert-Ergebnisse



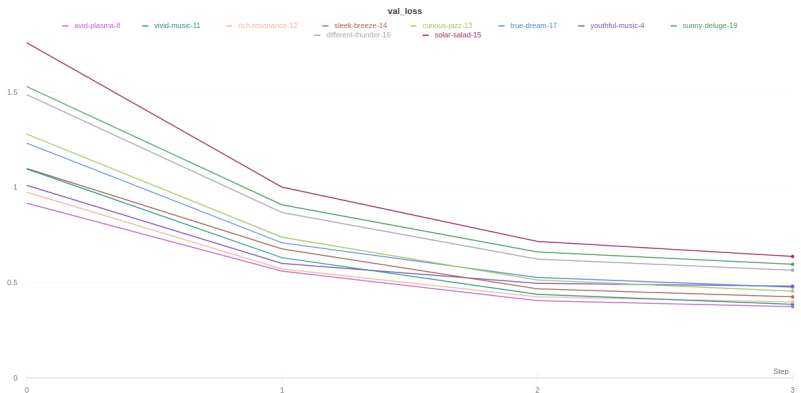
Beste Acc: 0.9084 mit Learning-Rate = 1e-4 und Batch-Size = 16

# Huggingface-Bert-Ergebnisse





# Huggingface-Bert-Ergebnisse



# Huggingface-Bert-Evaluation

sentence : I am <head>in</head> big trouble

prediction : indicating a state/condition/form, often a mental/emotional one that is being experienced

sentence : I am <head>in</head> a big airplane

prediction : indicating a LOCATION that surrounds (or can be viewed as surrounding) something else (e.g., in the morning)

sentence : I am <head>in</head> New York

prediction : indicating a LOCATION that surrounds (or can be viewed as surrounding) something else (e.g., in the morning)

sentence : <head>In</head> 2020 Donald Trump will be re-elected

prediction : indicating a TIME\_PERIOD during which something happens or continues (e.g., in the morning)

sentence : I always see you <head>in</head> my dreams.

prediction : indicating a LOCATION that surrounds (or can be viewed as surrounding) something else (e.g., in the morning)

sentence : I am speaking <head>in</head> portuguese.

prediction : indicating the language, medium, or means of encoding (e.g., spoke in German)

sentence : Donald Trump appears <head>in</head> a weird manner.

prediction : indicating a manner that something happens or is done, often somewhat idiomatic (e.g., in the morning)

# Huggingface-Bert-Evaluation

sentence : <head>By</head> 2021 I will have a bachelor degree.

prediction : indicating the size or amount, as of a margin (e.g., increase by 7%)

sentence : <head>By</head> the way, Joe Biden is actually better than is Opponent.

prediction : indicating the MEANS of achieving something (e.g., melt it by cooking it)

sentence : The Crisis was handled <head>by</head> the best president ever - Donald Trump.

prediction : indicating the logical subject (that is, the word that would be the subject in an act

sentence : He is leading the polls <head>by</head> far.

prediction : indicating the size or amount, as of a margin (e.g., increase by 7%)

# Huggingface-Bert-Evaluation-Details

preposition	accuracy	occurrence in test-data	different senses
through	0.6512	43	13
inside	0.7143	7	3
round	0.7500	12	3
around	0.7632	38	5
after	0.7778	9	6
before	0.8000	5	3
above	0.8000	5	5
beneath	0.8000	5	3
off	0.8125	16	4
like	0.8462	26	7
by	0.8462	52	8
in	0.8561	139	12
behind	0.8571	14	4
on	0.8750	88	19
during	0.8750	8	2
from	0.8852	122	16
into	0.8852	61	8
over	0.8947	19	8
between	0.9130	23	6
for	0.9263	95	11
of	0.9271	288	16
to	0.9322	118	10
down	0.9394	33	3
against	0.9474	19	6
at	0.9577	71	9
across	0.9688	32	2
with	0.9832	119	14
about	0.9859	71	4
as	1.0000	17	1
along	1.0000	36	3
towards	1.0000	22	4
onto	1.0000	10	2
beside	1.0000	6	1
among	1.0000	10	3

# Huggingface-Bert-Anbindung

```

<type5: Sentence begin="0" end="31" sofa="1" xmi:id="312"/>
<type5: Sentence begin="32" end="107" sofa="1" xmi:id="317"/>
<type6: WordSense begin="15" end="17" sofa="1" value="45" xmi:id="322"/>
<type6: WordSense begin="25" end="27" sofa="1" value="45" xmi:id="327"/>
<type6: WordSense begin="51" end="53" sofa="1" value="111" xmi:id="332"/>
<cas: Sofa mimeType="text" sofaID="_InitialView" sofaNum="1" sofaString="This is a test by Barack Obama. And this is another Test by the greatest president of the US - Donald Trump" xmi:id="1"/>

```

## FairNLP (Tim)

# FlairNLP

- Aktuell, spezialisiert für NLP-Aufgaben [Akbik, Blythe und Vollgraf 2018]
- Einfaches Framework
- Basiert auf PyTorch

# Flair Trainer

- Daten im CSV-Format in Korpus laden
  - Mindestens train.csv, optional dev.csv und test.csv
  - dev und test werden ggf. von flair erstellt
- Embeddings: FlairEmbeddings + DocumentRNNEndings
- Mindestens rund 100 Epochen Training



## Semi-Supervised (Tobias)

# Datenbeschaffung

- European Parliament Proceedings Parallel Corpus 1996-2011 (<http://www.statmt.org/europarl/>)
- CDEC Word Aligner (<https://github.com/redpony/cdec>)

# Daten Vorbereitung

- Tokenizen der Daten
- Alle Tokens in lower-case überführen
- Zusammenführen der beiden Corpora (special format)
- Entfernen von unvollständigen Zeilen
- Wörter alignen

# Verfolgte Ansätze

- Bidirectional LSTM selbst trainieren
- Transformers Model transfer learning
- Bert Transfer Learning

# Ausblick

- Verschiedene Sprachen ausprobieren
- Output in die anderen Classifier einbinden

## Quellen

# Quellen



Akbik, Alan, Duncan Blythe und Roland Vollgraf (2018).  
“Contextual String Embeddings for Sequence Labeling”. In:  
*COLING 2018, 27th International Conference on Computational  
Linguistics*, S. 1638–1649.