## Applied Deep Learning Homework 1

Kaggle Due: 2022/10/12 23:59

Code/Report Due: 2022/10/14 23:59

#### **UPDATES**

Sep 22 11:14 PM: Page 24 update default python version to 3.9

#### Links

- Sample Code
- Data
- Kaggle Intent Classification
- Kaggle Sequence Tagging

## Submission FAQ

- Reproduce有一點誤差可以嗎?
  - Reproduce 成功 = Reproduced 分數 >= Kaggle 分數 \* 0.99
  - Reproduce 成功 = Reproduced 分數 > Kaggle Baseline

#### TA Office Hour

- Thursday 10:00-11:30 (林彥廷)
- Wednesday 10:00-11:30 (郭蕙綺)
- Location: Online Google Meet

#### Outline

- Task Description
- Logistics
- Rules
- Report
- Guides
- Sample Code

Task Description

#### Part 1: Intent Classification

Input: Text
 "i dont like my current insurance plan and want a new one",
 "when will my american express credit card expire",
 "how would i get to city hall via bus",

#### • Output: Intent

```
"insurance_change",
"expiration_date",
"directions"
```

## Part 2: Slot Tagging

• Input: Text

"A table today for myself and 3 others"

"My three children and i are in the party"

• Output: Intent

"O O B-date O B-people I-people I-people O"

"B-people I-people I-people I-people O O O"

## Slot Tagging

- Slot Tagging: <u>Inside-Outside-Beginning tagging problem</u>
- Similar to NER task
- Classify each token in a sentence to a {O, B-xxx, I-xxx}
- After Preprocessing, this problem can be reduced to a multi-class classification problem

#### Metrics

- Intent Classification
  - Accuracy
- Slot Tagging
  - Joint Accuracy
  - A sample is correct only if all tokens are predicted correctly.

#### What to do

- Train an intent classification model and pass baselines:
  - Public Baseline: **0.86933**
  - Private Baseline: Released after deadline
- Train a slot tagging model and pass baselines:
  - Public Baseline: **0.71689**
  - Private Baseline: Released after deadline

#### Data

- Labeled data
  - o train.json
  - o eval.json
- Unlabeled data
  - o test.json
- Download link

## Data Format (json)

- Intent Classification
  - id: str
  - o text: str
  - Intent: str ☐ Only in train.json and eval.json
- Slot Tagging
  - id: str
  - o text: List[str]

## Field Description

- Intent Classification
  - o id: Unique id
  - text: Input sentence
  - o intent: A string that denotes the intent of the input sentence
- Slot Tagging
  - o id: Unique id
  - o text: A list of input tokens preprocessed from the input sentence
  - tags: A list of strings, each denotes the tag of its corresponding token in the input sentence

### Data example in train.json

#### Intent Classification

```
{
  "text": "send over a hundred dollars from huntington into saving",
  "intent": "transfer",
  "id": "train-110"
},
```

#### **Slot Tagging**

```
"tokens": [
 "a",
 "table",
 "for",
 "adults",
 "and",
 "4",
 "children",
 "please"
"tags": [
  "0",
  "0".
  "0".
 "B-people",
 "I-people",
 "I-people",
 "I-people",
  "I-people",
  "0"
"id": "train-3"
```

#### Submission Format - Intent Classification

- CSV (Comma Separated Values) format with 2 columns:
  - o id: Unique id for each sample
  - o intent: Your prediction. 要完全一樣不要多空白之類的

```
id,intent
test-0,todo_list_update
test-1,translate
test-2,insurance
```

## Submission Format - Slot Tagging

- CSV (Comma Separated Values) format with 2 columns:
  - o id: Unique id for each sample
  - tags: Your prediction. The tags should be separated with single space.

```
id,tags
test-0,0 0 0 B-people 0 0
test-1,0 0 0 0 0 0 0
test-2,0 B-first_name 0 0 0
```

## Logistics

## Grading

- Model Performance (10%)
  - Your intent classification model passes the baseline on the public test set
     (2%) and the private test set (3%) on kaggle
  - Your slot tagging model passes the baseline on the baseline on the public test set (2%) and the private test set (3%) on kaggle
  - Only if you can reproduce any submission that beats baseline in intent\_cls.sh slot\_tag.sh
- Format (1%)
  - TA can run the grading script without human intervention.
- Report (9% + 1% Bonus)
  - In PDF format!

## Code/Scripts/Report Submission

- Zip your folder into a single .zip file.
- Submit to NTU Cool.

## File Layout

Your zip must contain files (case sensitive):

- /[student id (lower-cased)]/, ex. /r12922000/, no brackets
  - o intent\_cls.sh
  - o slot\_tag.sh
  - o README.md
  - report.pdf
  - download.sh
  - Any other code/script.
- Do not upload training, validation, testing data and model to COOL.

#### Submission Files - download.sh

- download.sh to download your model. 可以用wget <url>
  - Do not modify your file after deadline, or it will be seen as cheating.
  - Keep the URLs in download.sh valid for at least 2 weeks after deadline.
  - On not do things more than downloading. Otherwise, your download.sh may be killed.
  - You can download at most 4G, and download.sh should finish within 1 hour.
- You can upload your model to <u>Dropbox</u>. (see <u>tutorial</u>)
- We will execute download.sh before predicting scripts.

## Submission Files - Scripts

- intent\_cls.sh, slot\_tag.sh
- Corresponding to the intent classification model, slot tagging model
  - "\${1}": path to the testing file. 路徑不要寫死!! 要考慮到電腦不一樣
  - "\${2}": path to the output predictions.
- TA will predict testing data as follow:
  - bash ./intent\_cls.sh /path/to/test.json /path/to/pred.csv
- Default python version would be 3.9 (Updated Sep. 22)
- Make sure your code works!

## Submission Files - Reproducibility

- All the code you used to train, predict, plot figures for the report should be upload.
- README.md 環境測試、資料處理等等
  - Write down how to train your model with your code/script **specifically**.
  - If necessary, you will be required to reproduce your results based on the README.md.
  - If you cannot reproduce your result, you may lose points.
- You will get at least 2 penalty if you have no or empty README.md.

#### **Execution Environment**

- Will be run on computer with
  - Ubuntu 20.04
  - o 32 GB RAM, GTX 3070 8G VRAM, 10G disk space available.
  - the packages we allow only.
  - o python 3.9
- Time limit <u>60</u> min for intent\_cls.sh slot\_tag.sh in total
- No network access when predicting.
- You will lose (some or all) your model performance score if
  - o your script is at wrong location, or cause any error.

## Rules

## Kaggle

- Displayed Team Name: [student\_id]
  - o e.g. r12345678
- For auditing, Displayed Team Name: audit\_[anything]
  - E.g. audit\_4fun
- You can submit your result 5 times a day for each task.
  - Any approaches to submit more than 5 times a day is prohibited!

#### What You Can Do

- Train with the data we give you.
- Use publicly available pre-trained word embeddings. (No contextualized word embedding.)

  (BERT, GPT等transformer相關的)
- Use the packages/tools we allow:
  - Python 3.9 and Python Standard Library
  - <u>PyTorch 1.12.1</u>, <u>TensorFlow 2.10.0</u>, <u>seqeval=1.2.2</u>
  - o tqdm, numpy, pandas, scikit-learn 1.1.2
  - Dependencies of above packages/tools.
- If you want to use other package, COOL/mail TA.

#### What You Can **NOT** Do

- Any means of cheating or plagiarism, including but not limited to:
  - Use others' code from anywhere (e.g. web, github, classmate, etc.).
  - Use the labels of the test data directly or indirectly. (Do not try to find them.)
  - Use package or tools not allowed.
  - Use model trained with other data.
  - Give/get model prediction to/from others.
  - Give/get trained model to/from others.
  - Publish your code before deadline.
- Violation may cause zero/negative score and punishment from school.

## **Submission Policy**

- Submit to NTU Cool.
- No Late submission.

## Report

# You may lose score if TA has difficulty understanding it.

Please write in a human-readable way.

## When Describing Model

- Please limit the use of imprecise words.
- Use equation whenever possible.
- Descriptions which is imprecise or hard to understand may cause loss of points.
- Ex.
  - bad: Feed the embedding of the sentence into a LSTM.
  - o good:  $h_t$ ,  $c_t = \text{LSTM}(w_t, h_{t-1}, c_{t-1})$ , where  $w_t$  is the word embedding of the t-th token.

## Q1: Data processing (2%)

- Describe how do you use the data for intent\_cls.sh, slot\_tag.sh:
  - a. How do you tokenize the data.
  - b. The pre-trained embedding you used.
- 2. If you use the sample code, you will need to explain what it does in your own ways to answer Q1.

## Q2: Describe your intent classification model. (2%)

- 1. Describe
  - a. your model
  - b. performance of your model.(public score on kaggle)
  - c. the loss function you used.
  - d. The optimization algorithm (e.g. Adam), learning rate and batch size.

## Q3: Describe your slot tagging model. (2%)

- 1. Describe
  - a. your model
  - b. performance of your model.(public score on kaggle)
  - c. the loss function you used.
  - d. The optimization algorithm (e.g. Adam), learning rate and batch size.

### Q4: Sequence Tagging Evaluation (2%)

- Please use <u>sequeval</u> to evaluate your model in Q3 on validation set and report *classification\_report(scheme=IOB2, mode='strict')*.
- Explain the differences between the evaluation method in <u>sequently</u>, token accuracy, and joint accuracy.

```
Ground Truth: [0 0 B-people I-people 0 0]
Prediction: [0 0 B-people B-people 0 0]

Ground Truth: [0 0 B-loc 0 0]
Prediction: [0 0 B-loc 0 0]

Joint Accuracy = 1 / 2
Token Accuracy = (5+5) / (6+5)
```

		precision	recall	f1-score	support
1	ISC	0.00	0.00	0.00	1
	PER	1.00	1.00	1.00	1
micro	avg	0.50	0.50	0.50	2
macro	avg	0.50	0.50	0.50	2
weighted	avg	0.50	0.50	0.50	2

## Q5: Compare with different configurations (1% + Bonus 1%)

- Please try to improve your baseline method (in Q2 or Q3) with different configuration (includes but not limited to different number of layers, hidden dimension, GRU/LSTM/RNN) and EXPLAIN how does this affects your performance / speed of convergence / ...
- Some possible BONUS tricks that you can try: multi-tasking, few-shot learning, zero-shot learning, CRF, CNN-BiLSTM
- This question will be grade by the completeness of your experiments and your findings.

## Guides

## Pipeline for (Deep) NLP

- Load pre-trained embedding (GloVe, ...).
- Preprocess the dataset
  - Tokenize the sentences (SpaCy).
  - Convert token to word indices.
- Prepare batch
  - Sample batch
  - Pad samples to the same length.
- Train, check metrics on validation.
- Predict!

Already written in the sample code

## Slot Tagging Pipeline

Tag 0 Tag 1 Tag N-1 Tag N Calculate loss with labels P(Slot=xxx) P(Slot=xxx) P(Slot=xxx) P(Slot=xxx) **Token Prediction** linear linear linear Linear Layer Recurrent Layer Input Token # tokens in text

## Sample Code

- Link
- TA will not explain the sample code for you.
- You can also write from scratch.
- Any bug report is welcome!
- See READMD.md for instructions