

# Applied Deep Learning Homework 1

Report/Code Due: 2021/04/11 23:59

# Links

[Homework Page](#)

[COOL Discussion](#)

[Sample Code](#)

[Data](#)

[Kaggle Intent Classification](#)

[Kaggle Sequence Tagging](#)

[adl-ta@csie.ntu.edu.tw](mailto:adl-ta@csie.ntu.edu.tw)

TA Hours:

- Mon 12:00-13:30 @ R524
- Thu 14:30-15:30 @ R524

# Change Log

March 15 18:21 - Launch Kaggle Competition

# 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 I-people* O O O O"

# Slot Tagging

- Slot Tagging: Inside-Outside-Beginning tagging problem
- Similar to NER task
- Classify each token in a sentence to a  $\{O, B\text{-xxx}, I\text{-xxx}\}$
- 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: **0.85022**
  - Private Baseline: **Released after deadline**
- Train a slot tagging model and pass baselines:
  - Public Baseline: **0.72171**
  - Private Baseline: **Released after deadline**

# Data


- Labeled data
  - train.json
  - eval.json
- Unlabeled data
  - test.json
- Check NTU Cool for download link.

# Data Format (json)

- Intent Classification

- id: str
- text: str
- Intent: str  *Only in train.json and eval.json*

- Slot Tagging

- id: str
- text: List[str]
- tags: List[str]  *Only in train.json and eval.json*

# Field Description (Intent Classification)

- Intent Classification
  - `id`: Unique id
  - `text`: Input sentence
  - `intent`: A string that denotes the intent of the input sentence
- Slot Tagging
  - `id`: Unique id
  - `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",  
    "2",  
    "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 - Slot Tagging

- CSV (Comma Separated Values) format with 2 columns:
  - 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
```

# Submission Format - Intent Classification

- CSV (Comma Separated Values) format with 2 columns:
  - id: Unique id for each sample
  - intent: Your prediction.

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



# 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 your best submission 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
  - `intent_cls.sh`
  - `slot_tag.sh`
  - `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.
  - 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.
  - Do 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 [Dropbox](#). (see [tutorial](#))
- 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
  - `"${1}":` path to the testing file.
  - `"${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`
- Specify the Python version (**only 3.8**) in the `.sh` file.
  - Ex. `python3.8 predict.py ...`
- **Make sure your code works!**

# Submission Files - Reproducibility

- All the code you used to train, predict, plot figures for the report should should be upload.
- README.md
  - 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
  - 32 GB RAM, GTX 3070 8G VRAM, 10G disk space available.
  - the packages we allow only.
  - python 3.8
- Time limit 60 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
  - your script is at wrong location, or cause any error.



# Rules

# Kaggle

- Link:
  - [Intent Classification](#) / [Slot Tagging](#)
- Displayed Team Name: [student\_id]
  - 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.)
- Use the packages/tools we allow:
  - [Python 3.8](#) and [Python Standard Library](#)
  - [PyTorch 1.7.1](#), [TensorFlow 2.4.1](#), [pytorch-lightning 1.2.3](#)
  - [SpaCy 3.0.5](#) for non-model-based functions.
  - [segeval=1.2.2](#), [tqdm](#)
  - 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.
- Late submission of "code and report":
  - 0 day < late submission  $\leq$  1 day: original score \* 0.95
  - 1 day < late submission  $\leq$  3 day: original score \* 0.90
  - 3 day < late submission  $\leq$  4 day: original score \* 0.75
  - 4 day < late submission  $\leq$  5 day: original score \* 0.50
  - 5 day < late submission  $\leq$  6 day: original score \* 0.25
  - 6 day < late submission: original score \* 0.00
- Late submission is determined by the last submission.
  - Update your submission after deadline implies that you will get penalty.

# 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.
  - 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%)

1. 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 [segeval](#) to evaluate your model in Q3 on validation set and report *classification\_report(schema=IOB2)*.
- Explain the differences between the evaluation method in [segeval](#), 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
MISC	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

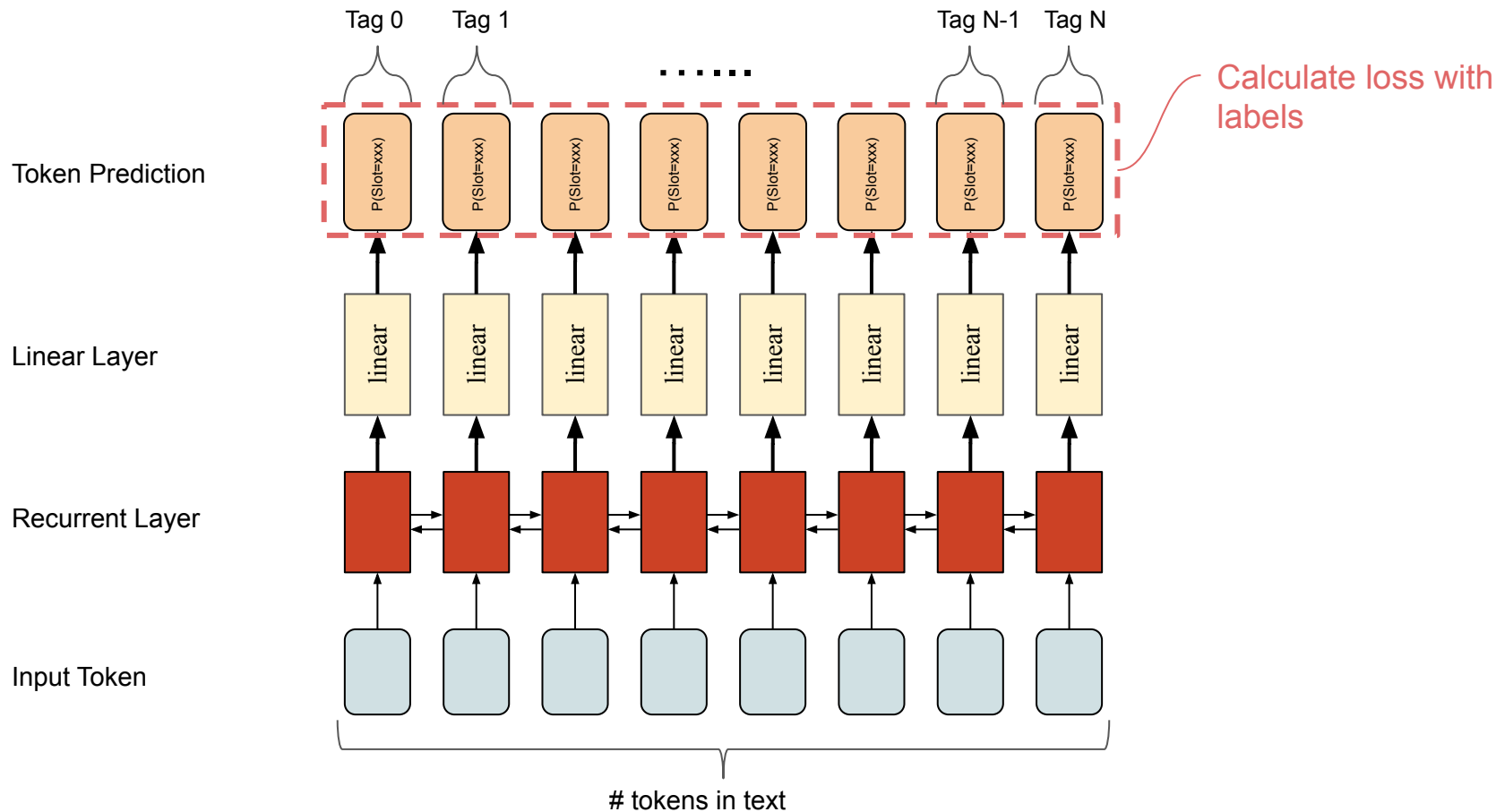
# Pipeline for (Deep) NLP

- 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





# Sample Code

- [Link](#)
- TA will not explain the sample code for you.
- You can also write from scratch.
- Any bug report is welcome!
- See README.md for instructions