

# NMT-GAN and EXPERIMENTS

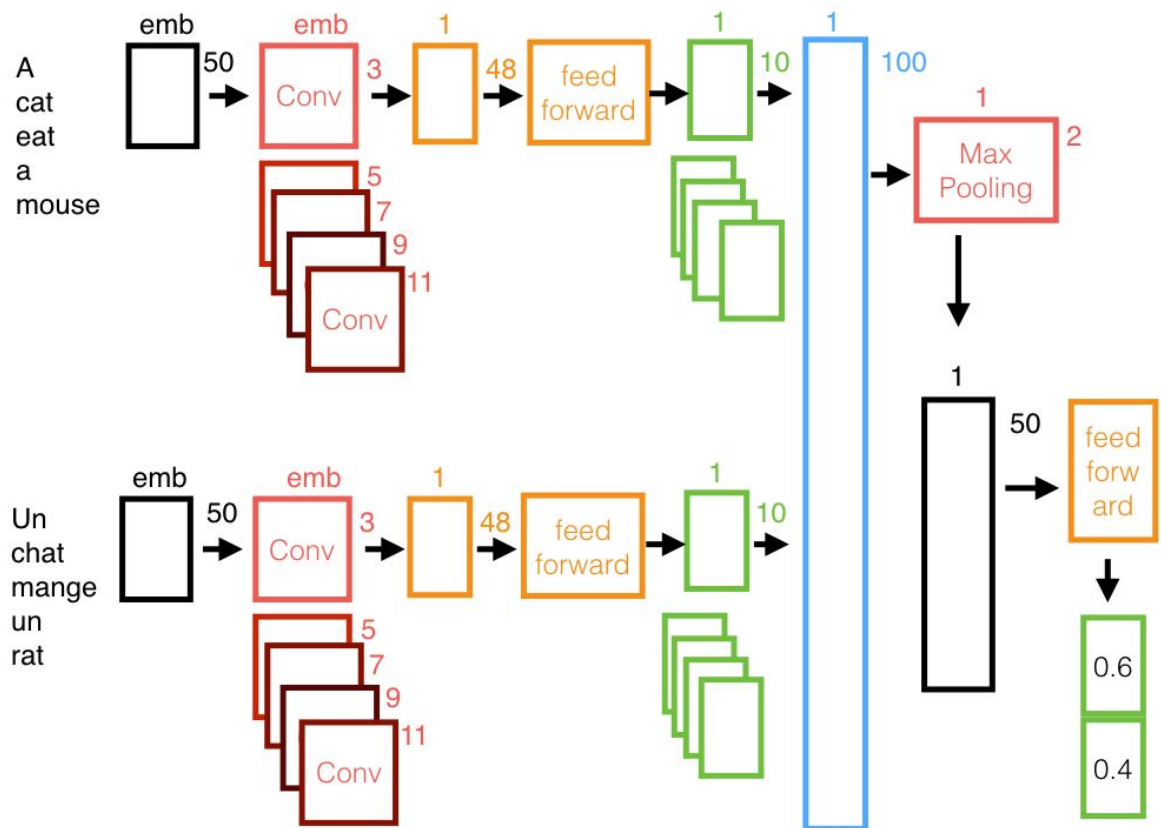
## Report 3

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## A. Problems

- a. CNN discriminator architecture: The sentence length is fixed into 50. The convolution size is a matrix of size 3, 5, 7, 9 and 11 to create different feature maps.



- b. Reward directly from Discriminator:

Its idea is from [Li et al]. The reward is assigned directly from Discriminator for each partially generated sentences.

In this case of training Discriminator, one example is randomly sampled from machine-generated and human-generated tokens for the negative and positive example respectively.

How to cut a sentence ?

- Example: [The reward is assigned directly from Discriminator for each partially generated sentences] = [The], [The reward], [The reward is], [The reward is assigned] ... until full sentence.

- c. Reward from French Language Model

The reward is from two components: Discriminator and Language Model. It is modified by a parameter alpha increasing for each batch.

The Language Model is trained with a larger dataset, based on the hypothesis that French has a large monolingual dataset but English.

- d. Controlling discriminator accuracy

There are two bounds: upper and lower bound.

```

    If lower bound < accuracy < upper bound:
best_discriminator = current_discriminator
    If accuracy < lower bound: Update discriminator
    If accuracy > upper bound: Use the best_discriminator

```

## B. Common settings

- a. Dataset:
  - i. GAN:
    - Train file : 19972 sentences
    - Valid file : 506 sentences
    - Test file : 469 sentences
  - ii. Language model:
    - Train file: 62227 sentences
    - Valid file: 31111 sentences
    - Test file: 31112 sentences
- b. Parameters:
- i. Batch\_size: 80
- c. Parameters for Generator:
  - i. Optimizer: Adadelata
  - ii. Loop of Generator: generator\_loop\_num: 1
  - iii. Embedding vector dimension:
    - embedding\_dim: 512
  - iv. RNN's hidden layer dimension:
    - rnn\_dim: 1024
  - v. Vocabulary:
    - n\_words\_src: 30000/ 7751
    - n\_words\_trg: 30000/ 9067
  - vi. Monte Carlo search:
    - maxlen: 50
    - Number of sentence rolled: rollnum: 20
- d. Parameters for Discriminator:
  - i. Optimizer: Adadelata
  - ii. Loop of Discriminator: discriminator\_loop\_num: 1
  - iii. Accuracy maximum: max\_acc: 0.85
  - iv. Accuracy minimum: min\_acc: 0.75
- e. Parameters for Language Model
  - i. Embedding vector dimension
    - in\_emb\_dim: 512
    - out\_emb\_dim: 512
  - ii. Vocabulary:
    - n\_words: 30000/ 7751
  - iii.  $\text{Alpha} * \text{Reward of Discriminator} + (1 - \text{Alpha}) * \text{Reward of LM}$

- $\alpha_{\text{init}} = 0$
- $\alpha_{\text{rate}} = 0.001$

## C. Algorithm

```
for each batch:
    alpha += alpha_rate
    for generator_loop_num:
        translated_sentences = translate_beam_search(batch)

        for sentence in translated_sentences:
            if monte_carlo_search:
                reward = get_reward_MC_sampling(sentence)
            else:
                reward = get_reward_not_MC(sentence)
        discriminator_rewards.append(reward)

        if language_model:
            for sentence in translated_sentences:
                reward = get_reward_LM(sentence)
                language_model_rewards.append(reward)

    rewards = alpha * discriminator_rewards + (1-alpha) *
language_model_rewards

    update_generator(translated_sentences, rewards)
    update_discriminator(batch, professor_rewards)

for discriminator_loop_num:
    if monte_carlo_search:
        data = prepare_date_MC_sampling(batch)
    else:
        data = prepare_date_not_MC(batch)
    update_discriminator(data)
```

## D. Experiments:

### a. NMT baseline

- Model type: Attention model with GRU
- BLEU score: (Beam\_search, argmax, multinomial)
  - Train: 46.71, 42.17, 12.98
  - Valid: 38.85, 35.11, 14.45
  - Test: 38.39, 36.06, 15.20

### b. NMT-GAN (MC search)

#### i. Pre-trained Generator:

- Model type: NMT baseline

#### ii. Pre-trained Discriminator:

- Model type: CNN discriminator
- Data: Machine-translated sentences of NMT baseline (Train 19972\*2 sentences, Valid 506\*2 sentences, Test 469\*2 sentences) with maximum length of sentence 50
- Optimizer: Adadelta
- Cross-entropy: (Multinomial\_1, Multinomial\_2)
  - Train: 0.125, 0.178
  - Valid: 0.141, 0.218
  - Test: 0.124, 0.177

#### iii. GAN:

- BLEU score:

### c. NMT with Reward directly from Discriminator

#### i. Pre-trained Discriminator: Same as NMT-GAN (MC search)

- Cross-entropy: (Multinomial)
  - Valid: 0.858, 0.835 (0.165)

#### ii. GAN:

- BLEU score:
  - Train: 46.71
  - Valid: 35.570
  - Test: 33.60

### d. NMT with Reward of Language Model

#### i. Language model: RNN

#### ii. GAN:

- BLEU score:

E. Code: <https://github.com/ngohoanhkhoa/GAN-NMT>

## F. Reference

- a. [Li et al]: [Adversarial Learning for Neural Dialogue Generation](#)