Deep Reinforcement Learning for Dialogue Generation

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Abstract

- Current neural network models.
 - They have drawbacks
- Solution:
 - Deep reinforcement learning.

Introduction

- Sequence-to-sequence (SEQ2SEQ) model (Sutskever et al., 2014).
- SEQ2SEQ is not without faults.

Introduction

- Drawbacks
 - Generate generic response
 - Can get stuck into infinite loop

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A: Where are you going? (1)
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B: I'm going to the restroom. (2)

A: See you later. (3)

B: See you later. (4)

A: See you later. (5)

B: See you later. (6)

...

...

A: Where are you going? (1)

B: I'm going to the police station. (2)

A: I'll come with you. (3)

B: No, no, no, no, you're not going anywhere. (4)

A: Why? (5)

B: I need you to stay here. (6)

A: I don't know what you are talking about. (7)

. . .

A: how old are you? (1)

B: I'm 16. (2)

A: 16? (3)

B: I don't know what you are talking about. (4)

A: You don't know what you are saying. (5)

B: I don't know what you are talking about . (6)

A: You don't know what you are saying. (7)

...

```
A: How old are you? (1)
```

B: I'm 16. Why are you asking? (2)

A I thought you were 12. (3)

B: What made you think so? (4)

A: I don't know what you are talking about. (5)

B: You don't know what you are saying. (6)

...

...

Introduction(cont...)

- What we need along with SEQ2SEQ
 - Developer-defined rewards.
 - Foresight.

Reinforcement Learning for Open-Domain Dialogue

- The learning system consists of two agents.
 p&q
 - Action
 - State
 - Policy

Reinforcement Learning for Open-Domain Dialogue

- Reward
 - R reward is obtained for each action.
 - Ease of answering
 - Information flow
 - Semantic coherence

- Simulate the process of two virtual agents.
 - Supervised learning
 - Mutual Information

- Mutual Information
 - Encoder-decoder model.
 - Stochastic gradient descent.

- Steps for dialogue simulation between two agents are-
 - Initiation step
 - Agent1 response
 - Agent2 response

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- Optimization
- Curriculum learning

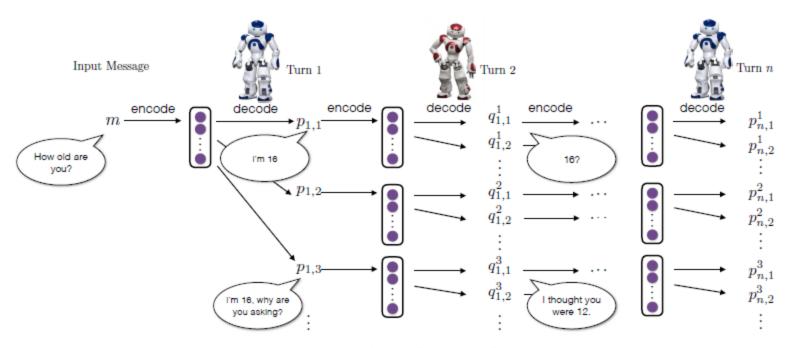


Figure 1: Dialogue simulation between the two agents.

- The dialogues are evaluated using two metrics
 - Conversation length
 - Diversity

- Data Set
 - High-quality initial inputs fed to the agent.
- Automatic Evaluation
 - Length of the dialogue
 - Diversity

- Human evaluation(3 settings)
 - Which is better
 - Easier to respond
 - High quality

Qualitative analysis

Input Message	Mutual Information Model
How old are you?	I'm 16.
What's your full name?	I have no idea.
I don't want to go home tonight.	Really?
Do you have any feelings for me?	I don't know what you are talking about.
How much time do you have here?	Not long enough. Sorry, sir.
Shall we get started?	Of course!
Do you play football?	No, i don't
We'd have to talk to him.	I mean, he's a good guy
How come you never say it?	Because I don't want to hurt you.

Qualitative analysis

Input Message	RL
How old are you?	I'm 16, why are you asking?
What's your full name?	What's yours?
I don't want to go home tonight.	Why?
Do you have any feelings for me?	I don't want to hurt your feelings.
How much time do you have here?	Ten seconds.
Shall we get started?	Yes. We've got a lot of work to do here.
Do you play football?	Yes. I love football!
We'd have to talk to him.	About what ?
How come you never say it?	I don't think it's a good idea to say it.

Issues observed

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A: What's your name?
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B: Daniel.

A: How old are you?

B. Twelve. What's your name?

A. Daniel.

B: How old are you?

A: Twelve. What's your name?

B: Daniel.

A: How old are you?

В ...

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Conclusion

- Overcame drawbacks of SEQ2SEQ
- Responses are comparable to RL