

Adicionando robustes semântica a sistemas de diálogo

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Sistemas de diálogo

criar um programa capaz de dialogar com ser humano

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MIND

A QUARTERLY REVIEW

OF

PSYCHOLOGY AND PHILOSOPHY

I.—COMPUTING MACHINERY AND INTELLIGENCE

By A. M. TURING

1. The Imitation Game.

I PROFOSE to consider the question, 'Can machines think?'
This should begin with definitions of the meaning of the terms
'machine' and 'think'. The definitions might be framed so as to
reflect so far as possible the normal use of the words, but this
attitude is dangerous. If the meaning of the words 'machine'
and 'think' are to be found by examining how they are commonly

Sistemas de diálogo

goal -driven vs non-goal driven

redes neurais

Sistemas de diálogo baseados em

Modelos de linguagem baseados em redes neurais

Nos chamamos de modelo de linguagem uma distribuição de probabildiade sobre uma sequencia de tokens em uma lingua natural.

$$P(x_1, x_2, x_3, x_4) = p$$

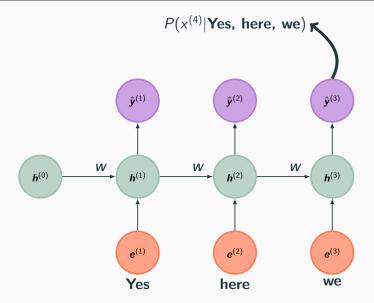
Em vez de usar uma abordagem que seja específica para o domínio da linguagem natural, podemos usar um modelo para predição de dados sequencias: **uma rede recorrente (RNN)**.

Nossa tarefa de aprendizado é estimar a distribuição de probabilidade

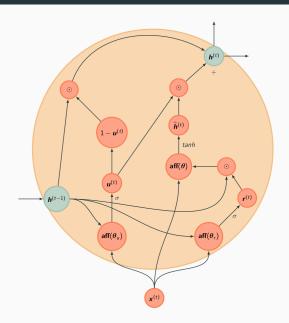
$$P(x_n = \text{palavra}_{j^*} | x_1, \dots, x_{n-1})$$

para qualquer (n-1) sequencia de palavras x_1, \ldots, x_{n-1} .

O modelo de linguagem com RNN



GRU: Gated Recurrent Units



Exemplo: TrumpBot

https://github.com/felipessalvatore/MyTwitterBot



Felipe Salvatore

Hillary can make america great again.

@greta @MarkBurnettTV #DinheiroNãoCompra #SecretBallot خسوف القم

Traduzir do inglês

15:10 - 7 de ago de 2017



Felipe Salvatore

@Felipessalvador

Obama is all beautiful. I agree with people attacking me. Amazing. @CLewandowski_#SecretBallot @garyplayer @greta

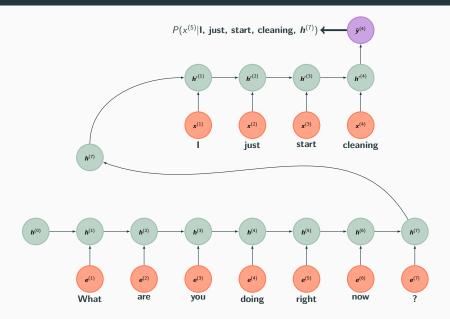
Traduzir do inglês

14:40 - 7 de ago de 2017

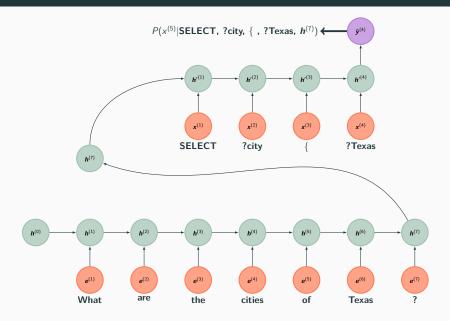
Exemplo: Funk Generator https://github.com/lucasmoura/funk_generator

É o di que tá tocando e não sabe de nada Eu já tô no clima e já tô no meu nome Cordão de ouro no pescoço eu tô na moda Com a camisa da Louis Vuitton Pulo da morena que elas gosta E se eu te pego no baile De captiva de citroen ou de hayabusa Tu viu a 1100 cilindradas Se eu tô no litoral de cordão de ouro De cordão de ouro no pescoço

Seq2seq: diálogo [5]



Seq2seq: tradução [3]



Exemplo de diálogo [5]

Human: what is moral?

Machine: what empowered humanity, what in-

tellectual the essence is . **Human:** what is immoral?

Machine: the fact that you have a child.

Human: what is morality? **Machine:** what is altruism?

Human: what is the definition of altruism?

Machine: if you don 't believe in god, then you

don 't know .

Human: *ok* ... *so what is the definition of morality* ?

Machine: well, the truth is, you 're not a believer in god almighty.

Human: tell me the definition of morality , i am

quite upset now!

Machine: i 'm not ashamed of being a philoso-

pher!

Métricas

Avaliação humana [2]

In the first trial, we asked the following questions to the users, for each response:

- 1. How appropriate is the response overall? (overall, scale of 1-5)
- 2. How on-topic is the response? (topicality, scale of 1-5)
- 3. How specific is the response to some context? (specificity, scale of 1-5)
- 4. How much background information is required to understand the context? (background, scale of 1-5)

Avaliação automática: BLEU (bilingual evaluation understudy) [4]

Essa métrica compara n-gramas (até 4) da resposta candidata com os n-gramas da refência da tradução e conta o numero de acertos. Essa métrica também penaliza traudções muito curtas:

$$BLUE(r,\hat{r}) = min\left(1, \frac{len(\hat{r})}{len(r)}\right) \left(\prod_{n=1}^{4} precision_n(r,\hat{r})\right)^{\frac{1}{4}}$$
(1)

em que $precision_n(r, \hat{r})$ é o número de overlap de n gramas de r e \hat{r} dividido pelo número de todos os n-gramas de \hat{r} .

$$BLUE(r,\hat{r}) \in [0,1]$$

12

Avaliação automática: problemas

wanna know
art>
all day and
A: @user i know, i 'm just so happy for you!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!

"In particular, we show that these metrics (BLEU, METEOR, ROUGE) have only a small positive correlation on the chitchat oriented Twitter dataset, and no correlation at all on the technical Ubuntu Dialogue Corpus." [1]

De diálogos abertos para

pequenas tarefas

bAbl [6]

Criar uma série de pequenas tarefas para testar diferentes capacidades de um sistema de diálogo.

Task 1: Single Supporting Fact

Mary went to the bathroom.

John moved to the hallway.

Mary travelled to the office.

Where is Mary? A:office

Task 3: Three Supporting Facts

John picked up the apple.

John went to the office.

John went to the kitchen.

John dropped the apple.

Where was the apple before the kitchen? A:office

Task 5: Three Argument Relations

Mary gave the cake to Fred.

Fred gave the cake to Bill.

Jeff was given the milk by Bill.

Jeff was given the milk by Bill.

Who gave the cake to Fred? A: Mary

Who did Fred give the cake to? A: Bill

Task 2: Two Supporting Facts

John is in the playground.

John picked up the football.

Bob went to the kitchen.

Where is the football? A:playground

Task 4: Two Argument Relations

The office is north of the bedroom.

The bedroom is north of the bathroom.

The kitchen is west of the garden.

What is north of the bedroom? A: office

What is the bedroom north of? A: bathroom

Task 6: Yes/No Ouestions

John moved to the playground.

Daniel went to the bathroom. John went back to the hallway.

Is John in the playground? A:no

Is Daniel in the bathroom? A:ves

is Daniel in the bathroom? A:yes

modelo de atenção

modelo de memória

ParlAI

https://github.com/facebookresearch/ParlAI

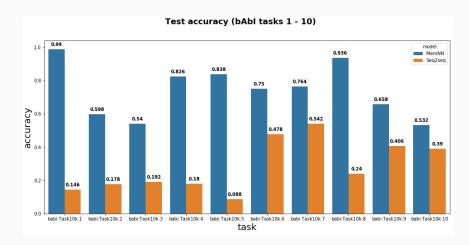


"ParlAI (pronounced 'par-lay') is a framework for dialog AI research, implemented in Python.

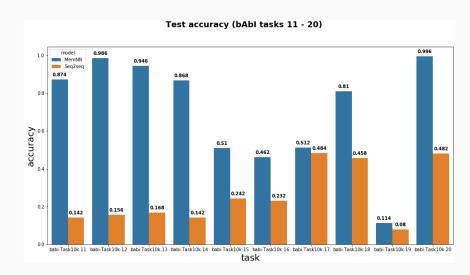
Its goal is to provide researchers:

- a unified framework for sharing, training and testing dialog models
- many popular datasets available all in one place, with the ability to multi-task over them
- seamless integration of Amazon Mechanical Turk for data collection and human evaluation"

Experimentos



Experimentos



Entailment-QA

bAbl: task 15

Basic Deduction

Task 15: Basic Deduction

Sheep are afraid of wolves.
Cats are afraid of dogs.
Mice are afraid of cats.
Gertrude is a sheep.
What is Gertrude afraid of? A:wolves

 P^1 are afraid of Q^1 P^2 are afraid of Q^2 P^3 are afraid of Q^3 P^4 are afraid of Q^4 c^1 is a P^1 c^2 is a P^2 c^3 is a P^3 c^4 is a P^4 What is c^j afraid of?

bAbl: task 16

Basic Induction

Task 16: Basic Induction

Lily is a swan.

Lily is white.

Bernhard is green.

Greg is a swan.

What color is Greg? A:white

$$c^1$$
 is a P^1
 c^1 is C^1
 c^2 is a P^2
 c^2 is C^2
 c^3 is a C^3
 c^4 is a C^4
 c^4 is C^4
 c^4 is a C^4
 c^6
 c^7
 c^8

SICK

quora

DialogGym



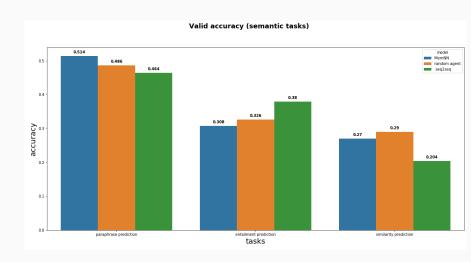
DialogGYM

https://github.com/felipessalvatore/DialogGym

Um novo conjunto de tarefas

- Task 1: entailment prediction Given two sentences *p* and *q* the agent is asked to detect a basic entailment relation between them, i.e., the agent should respond if *p* implies *q*, if *p* contradicts *q* or if *p* is neutral to *q*. For example, the sentences "A man is thinking" and "There is no man thinking" is given to the agent, he needs to detect the quantifier to spot the contradiction between these two informations.
- Task 2: similarity prediction The agent is questioned to indicate how related are the meaning of two sentences, e.g., "A man is reading the email. Someone is reading the email. Are the sentences above related?". There are only 4 possible answers: "not related", "somewhat related", "related", "strongly related".
- Task 3: paraphrase prediction The agent is asked (a yes/no question) to identify if two given questions express the same meaning using different words, e.g., "Who was Pele? Who is Pele? Are the above questions duplicate?".

Primeiros resultados



SICK como melhorar

Entailment-QA

comentar o sick

Entailment-QA

- 1. Boolean Connectives
- 2. First-Order Quantifiers
- 3. **Synonymy**
- 4. Antinomy
- 5. **Hypernymy**
- 6. Active/Passive voice

- Entailment $(s_1 \text{ implies } s_2)$
 - $P^1a^1 \wedge \cdots \wedge P^na^n$, P^ja^j
- Not entailment (s_1 does not implies s_2)

 - $\underbrace{P^{j}a^{j}}_{s_{1}}, \underbrace{P^{1}a^{1} \wedge \cdots \wedge P^{n}a^{n}}_{s_{2}}$ $\underbrace{P^{1}a^{1} \vee \cdots \vee P^{n}a^{n}}_{s_{1}}, \underbrace{P^{j}a^{j}}_{s_{2}}$

Ashley is fit

Ashley is not fit

The first sentence implies the second sentence? A: no

Avery is nice and Avery is obedient

Avery is nice

The first sentence implies the second sentence? A: yes

Elbert is handsome or Elbert is long

Elbert is handsome

The first sentence implies the second sentence? A: no

- Entailment
 - ∀xPx, Pa
 - Pa, ∃xPx
- Contradiction
 - $\forall x P x, \neg P a$
 - $\forall x P x, \exists x \neg P x$
- Neutral
 - Pa, Qa
 - $\forall x P x, \neg Q a$

Every person is lively

Belden is lively

What is the semantic relation? A: entailment

Every person is short

There is one person that is not short

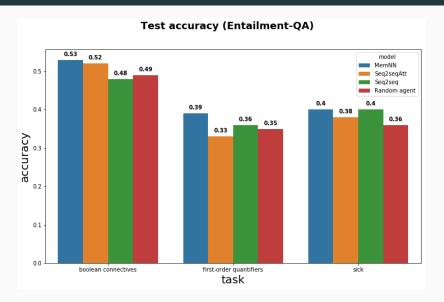
What is the semantic relation? A: contradiction

Every person is beautiful

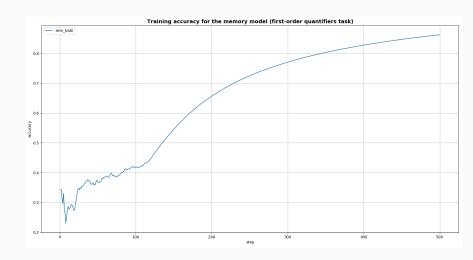
Abilene is not blue

What is the semantic relation? A: neutral

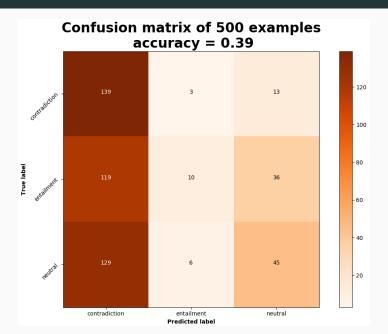
Resultados até agora



Resultados até agora



Resultados até agora



Próximos passos

- Terminar as tarefas
- Melhor o treinamento com os modelos atuais
- Explorar novos modelos

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