

## EMNLP 2021 Tutorial

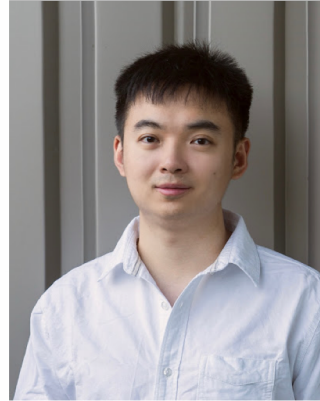
# Knowledge-Enriched Natural Language Generation



Wenhao Yu<sup>1</sup>,



Meng Jiang<sup>1</sup>,



Zhiting Hu<sup>2</sup>,



Qingyun Wang<sup>3</sup>,



Heng Ji<sup>3,4</sup>,



Nazneen Rajani<sup>5</sup>

1 University of Notre Dame    2 University of California San Diego

3 University of Illinois at Urbana-Champaign    4 Amazon Scholar    5 Salesforce Research

# Knowledge-Enriched NLG tutorial (Outline)

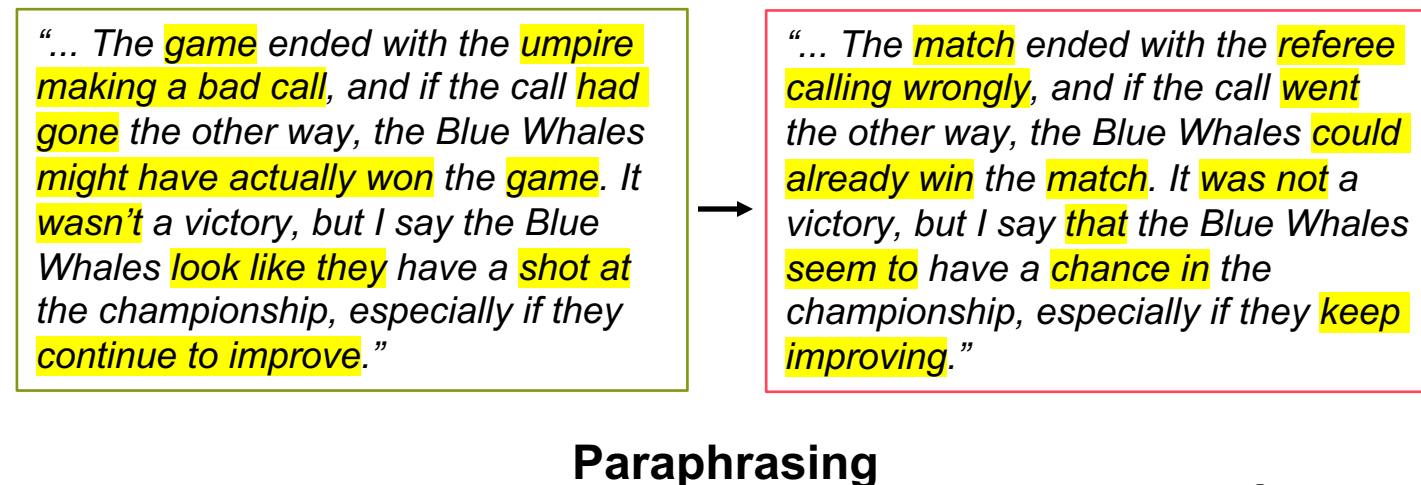
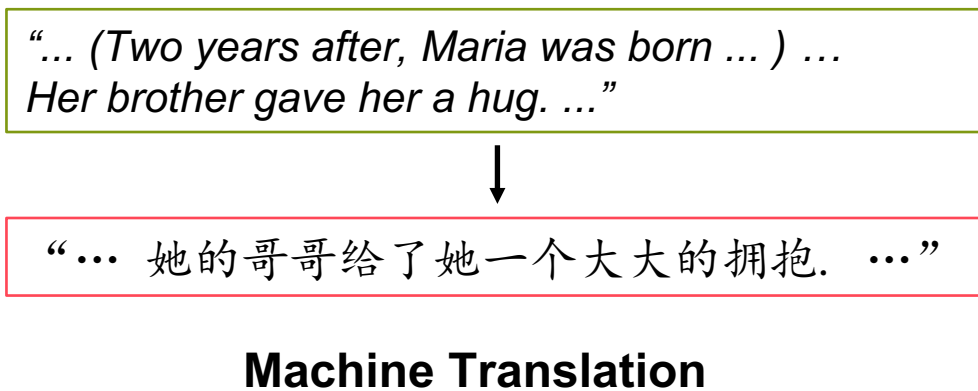
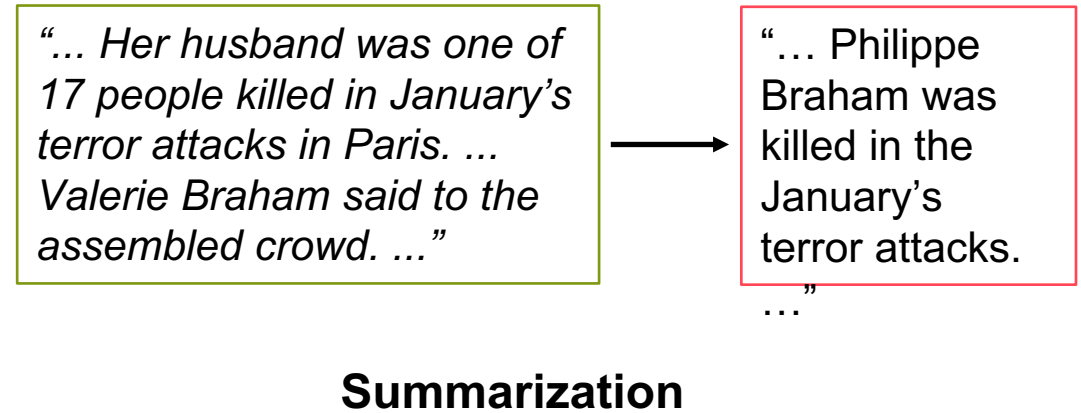
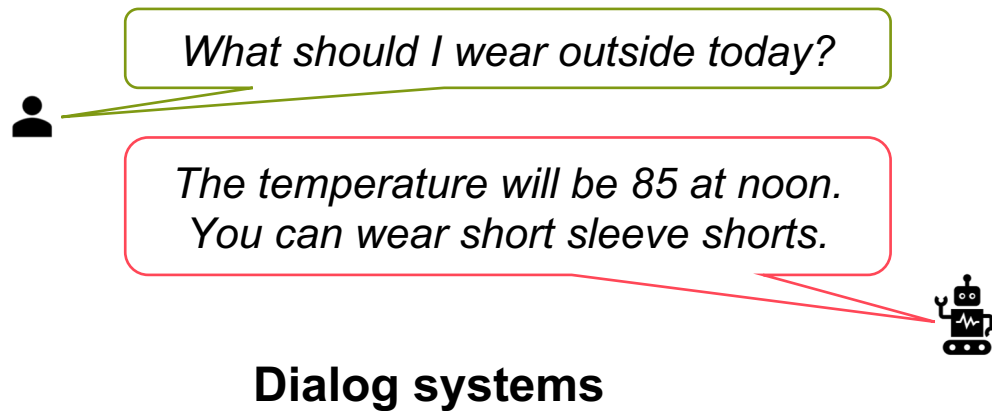


| Time (AST / EST)                             | Content   | Tutor                         |
|--|---|-------------------------------|
| 09:00 am – 09:10 am<br>(08:00 am – 08:10 am) | Motivation and Key Questions                      | Meng Jiang                    |
| 09:10 am – 09:50 am<br>(08:10 am – 08:50 am) | General Methods of Integrating Knowledge into NLG | Zhiting Hu,<br>Nazneen Rajani |
| 09:50 am – 10:15 am<br>(08:50 am – 09:15 am) | NLG Methods Enhanced by Various Knowledge Source  | Wenhao Yu                     |
| 10:15 am – 10:45 am<br>(09:15 am – 09:45 am) | Coffee break                                      |                               |
| 10:45 am – 11:10 am<br>(09:45 am – 10:10 am) | NLG Methods Enhanced by Various Knowledge Source  | (cont.)                       |
| 11:10 am – 12:10 pm<br>(10:10 am – 11:10 am) | Applications, Benchmark Datasets, Code Library    | Qingyun Wang                  |
| 12:10 pm – 12:30 pm<br>(11:10 am – 11:30 am) | Remaining Challenges and Future Directions        | Heng Ji                       |

# Text generation is needed everywhere in our life!



- Leveraging machine intelligence to make many things EASY and FAST



# Text generation is needed everywhere in our life!



**Image Captioning:**  
**Image-to-Text Generation**

|                         | 85     | Final | 95   |     |
|-------------------------|--------|-------|------|-----|
| ROCKETS                 | LAKERS | STATS | NEWS |     |
| Player                  | Min    | Reb   | Ast  | Pts |
| 6 LeBron James · F      | 35     | 7     | 8    | 15  |
| 9 Kent Bazemore · F     | 27     | 2     | 1    | 9   |
| 3 Anthony Davis · C     | 33     | 13    | 2    | 16  |
| 0 Russell Westbrook · G | 35     | 8     | 9    | 20  |
| 20 Avery Bradley · G    | 30     | 3     | 0    | 2   |
| 7 Carmelo Anthony       | 25     | 3     | 0    | 23  |
| 15 Austin Reaves        | 20     | 2     | 0    | 2   |
| 11 Malik Monk           | 19     | 2     | 0    | 0   |
| 10 DeAndre Jordan       | 17     | 3     | 2    | 8   |

HOU 1-5 85 Final 95 LAL 4-3

LOS ANGELES -- — [Carmelo Anthony](#) scored 23 points in a reserve role and [Russell Westbrook](#) added 20 points in the Los Angeles Lakers' 95-85 victory over the [Houston Rockets](#) on Sunday night.

[Anthony Davis](#) had 16 points and 13 rebounds, and [LeBron James](#) scored 15 points in the Lakers' fourth win in five games after opening the season with two losses. Westbrook added eight rebounds and nine assists.

[Eric Gordon](#) scored 17 points and [Christian Wood](#) had 16 for the Rockets, who lost their fourth straight at the start of a five-game road trip.

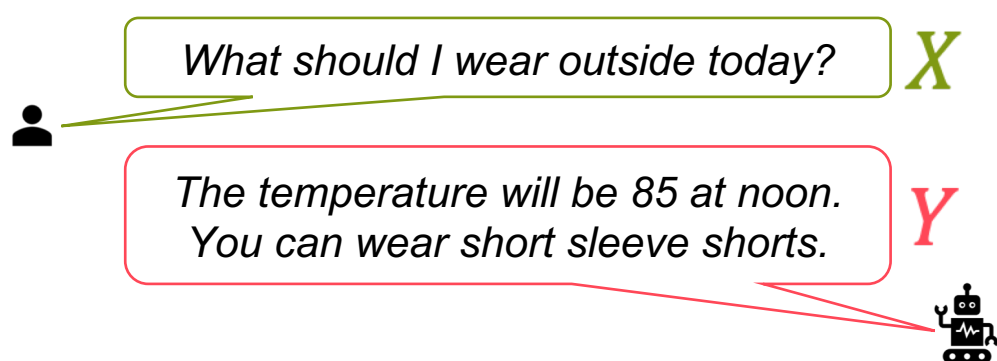
**Sports News Generation:**  
**Table-to-Text Generation**

*Not the main focus of this tutorial.*

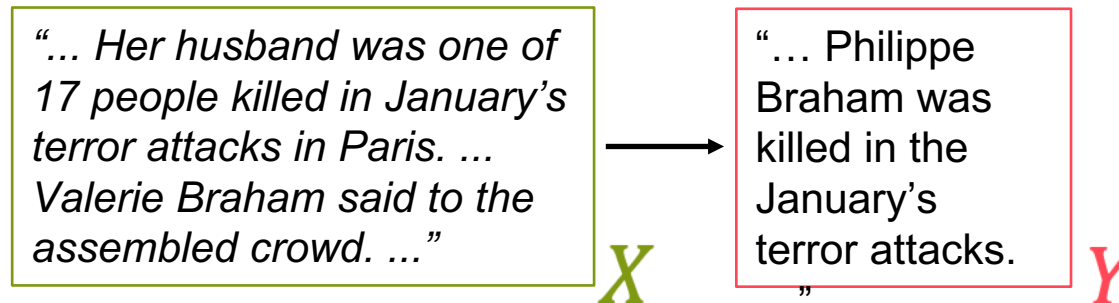
# Text-to-text generation is needed everywhere!



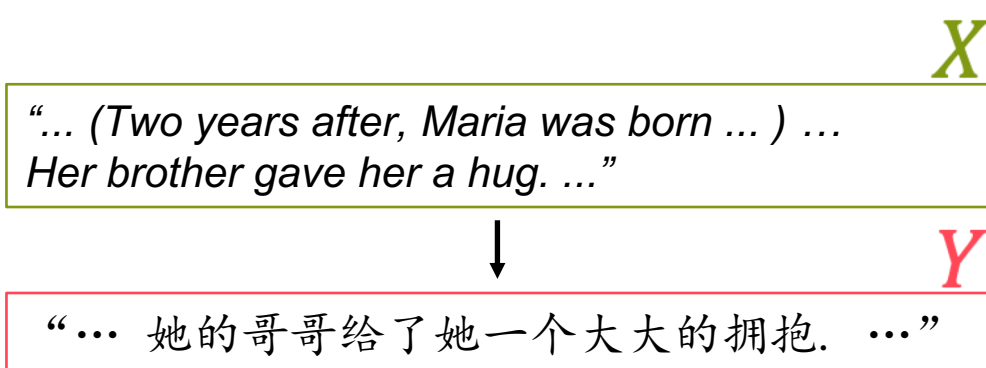
- $P(Y|X) = P(y_1, \dots, y_m | x_1, \dots, x_n) = \prod_{t=1}^m p(y_t | X, y_1, \dots, y_{t-1})$ , when **Y is text** and **X is text**.



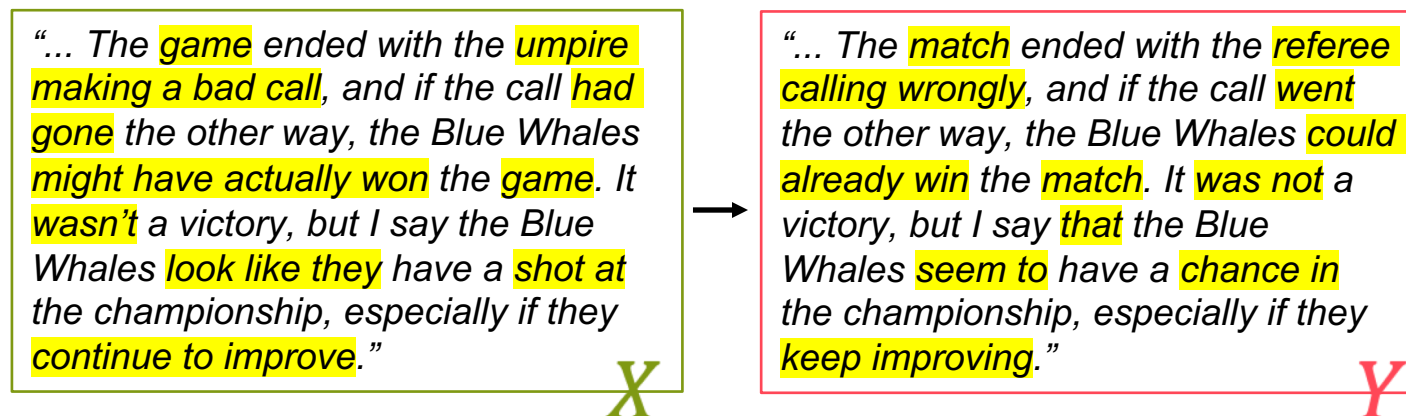
Dialog systems



Summarization



Machine Translation



Paraphrasing

# Knowledge is needed everywhere in text generation!



- Mistakes may occur due to lack of knowledge.

What should I wear outside today?

The temperature will be ~~65~~ 85 at noon.  
You can wear short sleeve shorts.

## Dialog systems

“... Her husband was one of 17 people killed in January’s terror attacks in Paris. ... Valerie Braham said to the assembled crowd. ...”

“... ~~Valerie Braham~~  
Philippe Braham was killed in the January’s terror attacks. ...”

## Summarization

“... (Two years after, Maria was born ... ) ... Her brother gave her a hug. ...”

“... 她的 ~~弟弟哥哥~~ 给了她一个大大的拥抱. ...”

## Machine Translation

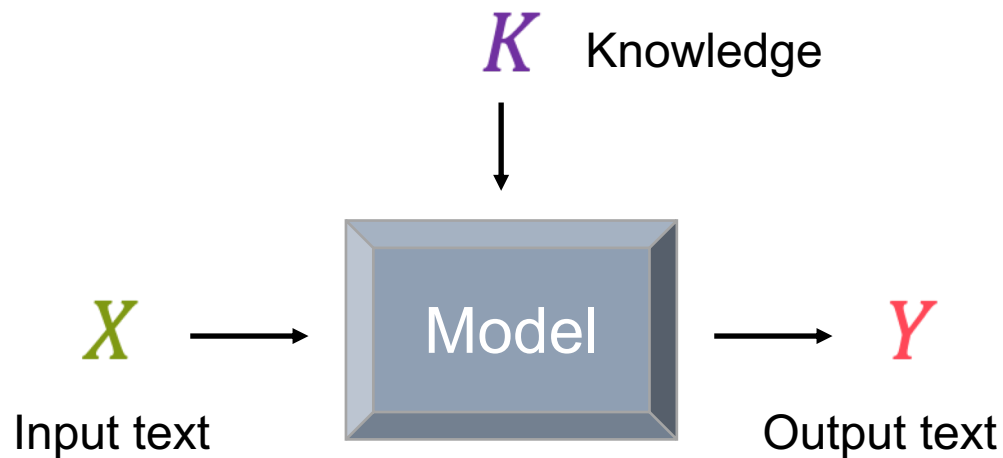
“... The **game** ended with the **umpire making a bad call**, and if the call **had gone** the other way, the Blue Whales **might have actually won** the **game**. It wasn’t a victory, but I say the Blue Whales **look like they** have a **shot at the championship**, especially if they **continue to improve**.”

“... The **match** ended with the **referee calling wrongly**, and if the call **went** the other way, the Blue Whales could **never** already win the **match**. It **was not** a victory, but I say **that the Blue Whales seem to** have a **chance in** the championship, especially if they **keep improving**.”

## Paraphrasing

# Knowledge-Enriched Natural Language Generation (KE-NLG)

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Key questions: (Outline of this tutorial)

- **Where** does the Knowledge come from? In other words, what are the **Knowledge sources**?
- **What** are the **data/representations** of Knowledge?
- **What** are the **methods** for integrating Knowledge into NLG models? **How** to utilize different types of Knowledge?
- **What** are the concrete NLG **models/techniques** that have been enhanced by Knowledge? **What** are their **advantages** and **disadvantages**?
- **Where** can we find **benchmark datasets, code library, and hands-on tutorial** for doing research on this topic?
- **What** are the remaining **challenges** and **future directions**?

# Material for this tutorial

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- Website: <https://kenlg-tutorial.github.io/>
  - Schedule
  - Slides
  - Tutor information
- A pre-print survey on the topic of Knowledge-enhanced NLG:
  - <https://arxiv.org/abs/2010.04389>
- A reading list and a collection of open-source code:
  - <https://github.com/wyu97/KENLG-Reading>

