Conditional Generation with a Question-Answering Blueprint

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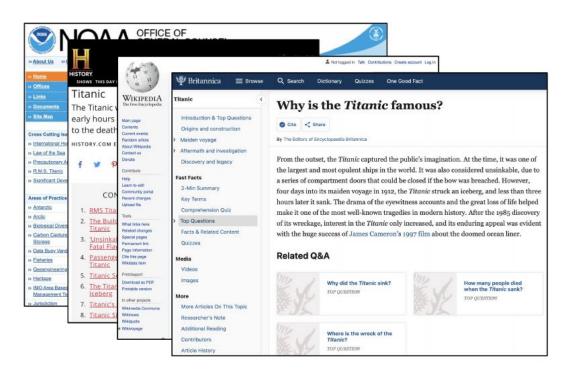


Conditional Generation: The Wish List

- Generate natural language towards a communicative goal
- which is faithful and can be attributed to is sources
- while users explicitly control generation outcome
- without being sensitive to the modality of the input

Conditional Generation: Long-form QA

Q what is the titanic known for



- Some queries have only long-form answers
- based on input from multiple documents
- where hallucinations and attribution can be more problematic

Conditional Generation: Dialogue Summarization

The Justice League Recombination, The Big Bang Theory, S4.E11



Penny starts dating Zack again. Used to being the target of bullying, Leonard, Sheldon, Howard and Raj instead bully Zack through their implications to him that he's stupid. Not fully realizing what they're doing until Penny points it out to them, the guys decide to apologize to Zack in the only way they can. Accepting their apology, Zack decides to start hanging out with them, much to Penny's chagrin. This hanging out leads to the idea that Zack should replace Leonard as Superman in their group Justice League of America costume for the New Year's Eve party at the comic book store, with Penny the sixth wheel as Wonder Woman. These actions lead to questions from Leonard to Penny about why she's back together with Zack.

Conditional Generation: Visual Storytelling

Image Sequence:











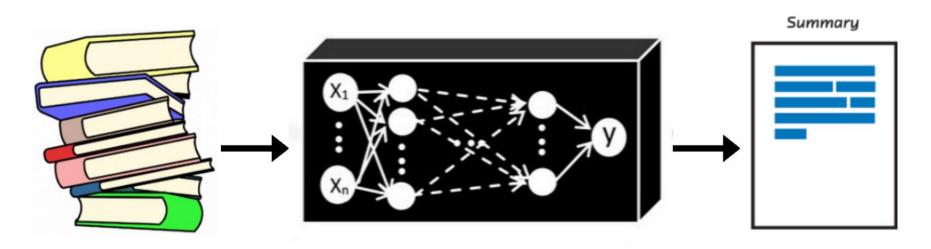
Gold-standard Story:

It was customary to give big brass a full tour. Of course, the mess hall would usually serve something a little better than average on "Big Brass" days. He looked like he was okay with it. After lunch the tour continued. He was introduced to the Commanding officers.

Story Generated by KE-VIST:

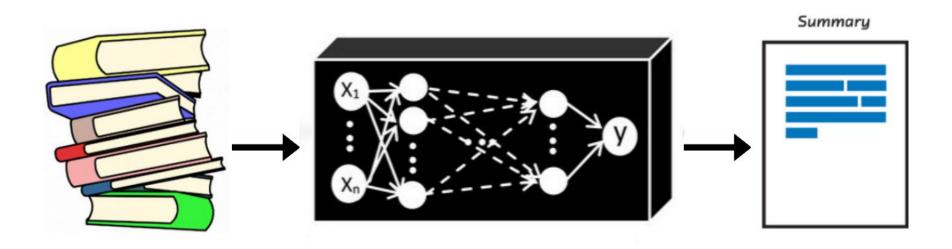
We all met up with the men. They walked a lot in the man walking. We had a lot of food. I gave a

The Blackbox Nature of Conditional Generation



- Change the way entities are represented (Puduppully et al., 2019; Iso et al., 2019)
- The decoder skips low-confidence tokens (Tian et al., 2019)
- Encode documents hierarchically (Rhode et al., 2021)
- Adopt sparse attention mechanisms (Child et al., 2019; Beltagy et al., 2020)

The Blackbox Nature of Conditional Generation



Introduce intermediate planning components (Wiseman et al., 2018; Moryossef et al., 2019; Puduppully et al., 2022; Narayan et al., 2022)

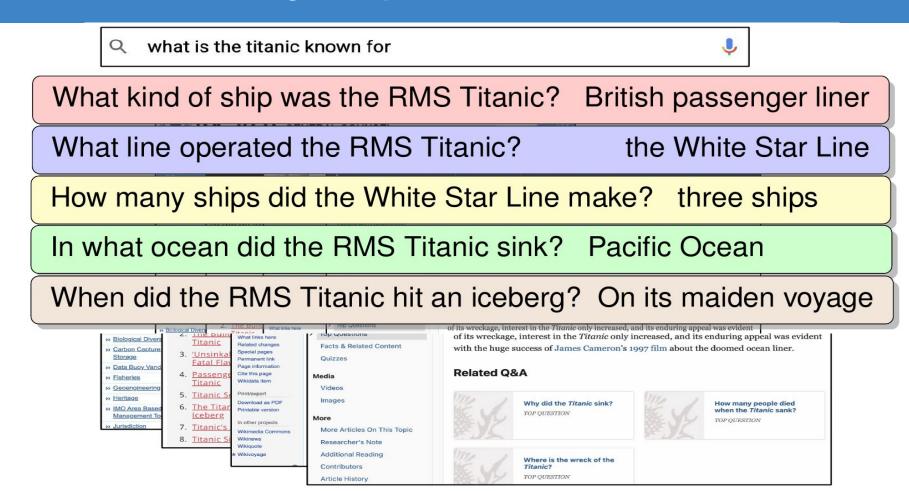
The QUD Model of Discourse

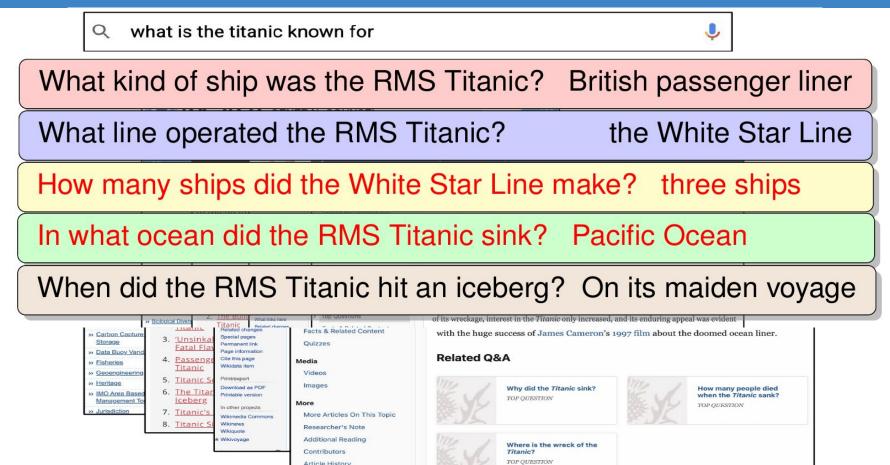
Questions under Discussion (QUD) theory of discourse structure (Carlson, 1983; Ginzburg, 1994; Van Kuppevelt, 1995; Larson, 2002; Roberts, 2012; Riester, 2019).

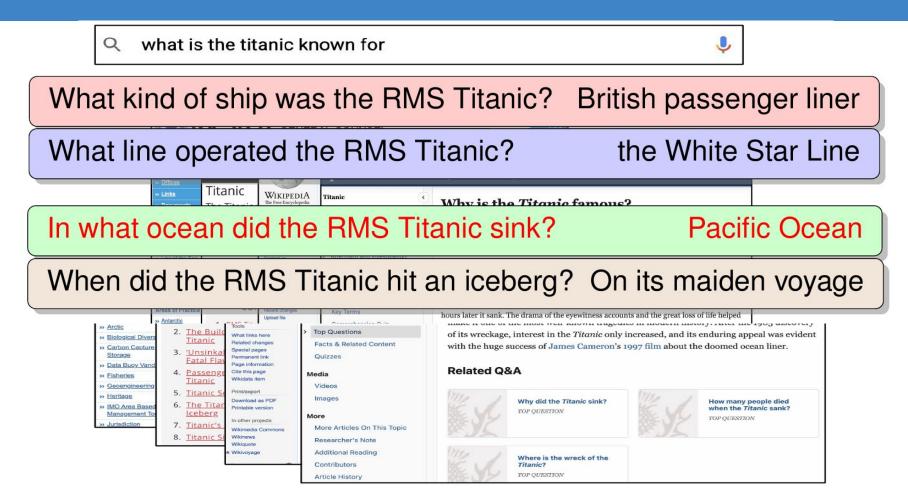
A **QUD** is a partially structured set of questions which discourse participants are mutually committed to resolving at a given point in time.

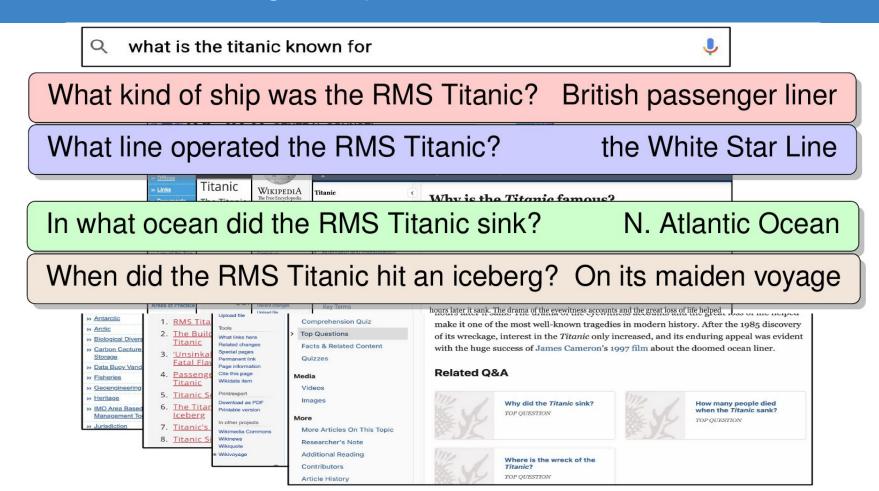
A **question** is a set of alternative possibilities (corresponding in some way to the possible answers to the question).

Discourse contains **implicit questions** for each of the assertions made which are turned into answers.









Q what is the titanic known for



What kind of ship was the RMS Titanic? British passenger liner

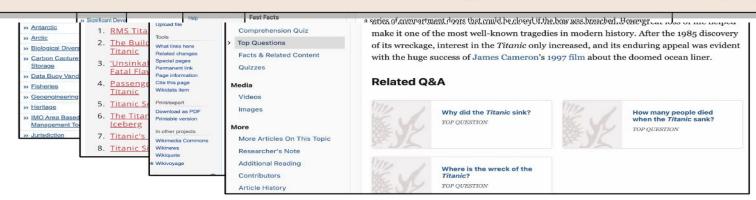
What line operated the RMS Titanic?

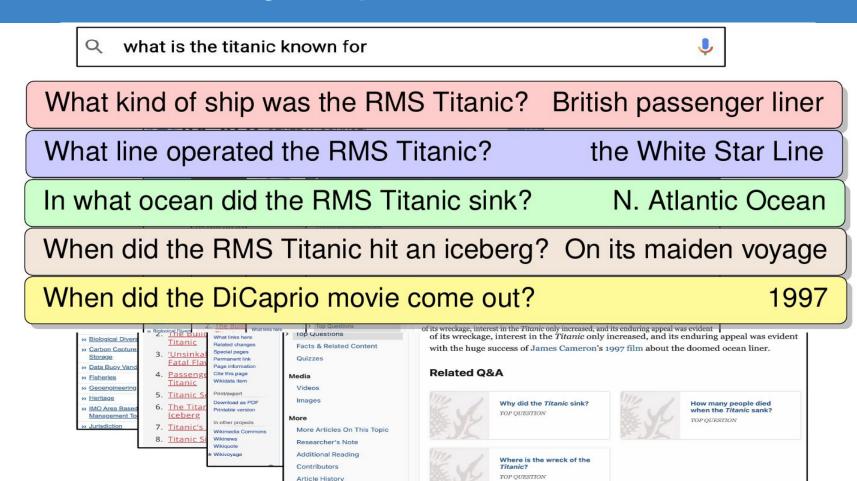
the White Star Line

In what ocean did the RMS Titanic sink?

N. Atlantic Ocean

When did the RMS Titanic hit an iceberg? On its maiden voyage





what is the titanic known for What kind of ship was the RMS Titanic? British passenger liner What line operated the RMS Titanic? the White Star Line In what ocean did the RMS Titanic sink? N. Atlantic Ocean When did the RMS Titanic hit an iceberg? On its maiden voyage When did the DiCaprio movie come out? 1997

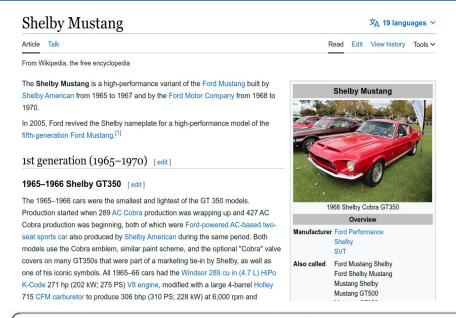
RMS Titanic was a British passenger liner, operated by the White Star Line. It sank in the North Atlantic Ocean after striking an iceberg during its maiden voyage. The 1997 film Titanic starring Leonardo Di Caprio has popularized the disaster and its aftermath.

- Blueprints as intermediate (discrete) planning stage for conditional generation
- Reduce faithfulness errors and increase controllability
- Are better for long-form inputs and outputs

Where do the blueprints come from?

Datasets as we know them consist of document(s)-output pairs!

Training Data Annotation with Blueprints



Training Data Annotation with Blueprints

| What is a high performance variant of the Ford Mustang? | Shelby |
|--|--------------------|
| Who built the Shelby Mustang from 1969 to 1970? | Ford |
| Q2: What is the high performance variant of the Ford Mustang called? | Shelby |
| What is a high performance variant of the Ford Mustang? | Shelby Mustang |
| What is a high performance variant of the Ford Mustang? | Shelby |
| What is a Shelby Mustang? A high per | rformance variant |
| The Shelby Mustang is a high performance variant of what? | Ford Mustang |
| During what years was the Shelby Mustang built by Shelby American? | ? 1965 to 1968 |
| Who built the Shelby Mustang from 1965 to 1968? | Shelby American |
| In what year was the fifth generation of the for Mustang introduced? | 2005 |
| What generation of Mustang was introduced in 2005? | e fifth generation |
| What was the Shelby Mustang revived as? a new high-performance model | |

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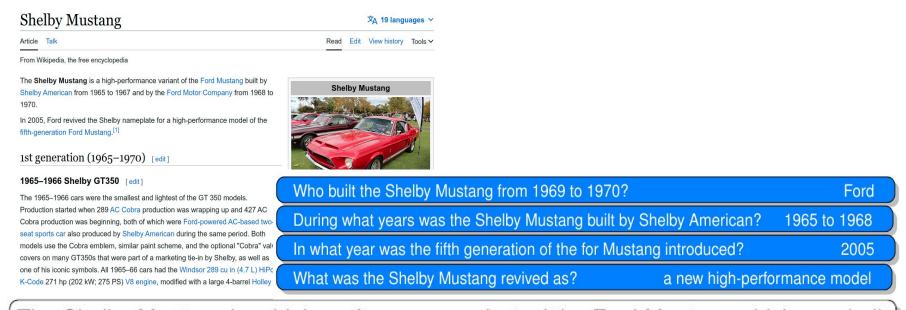
The Shelby Mustang is a high performance variant of the Ford Mustang which was built by Shelby American from 1965 to 1968, and from 1969 to 1970 by Ford. Following the introduction of the fifth generation Ford Mustang in 2005, the Shelby nameplate was revived as a new high-performance model, this time designed and built by Ford.

Who built the Shelby Mustang from 1969 to 1970?

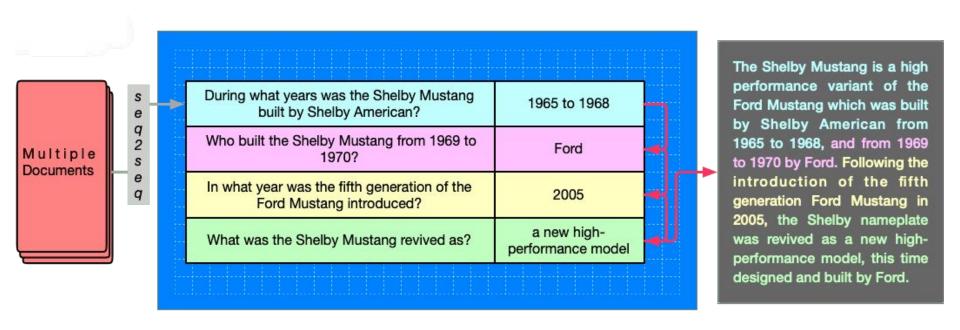
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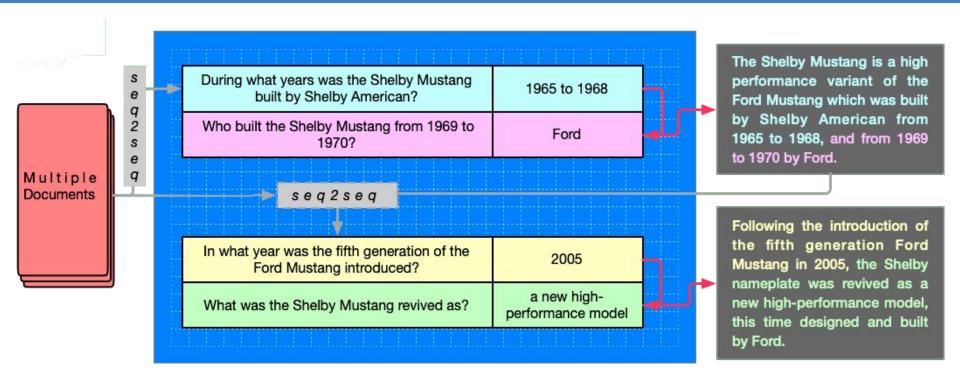


End-to-End Blueprint Model



Generates blueprint and output sequence in one go.

Iterative Blueprint Model



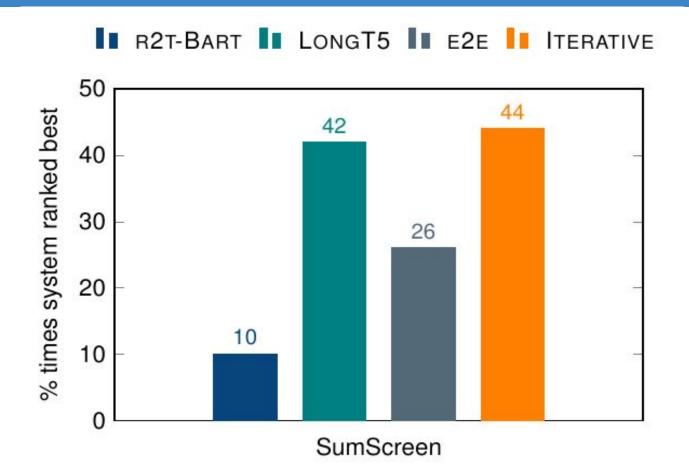
Iteratively plans and generates one sentence at a time.

Experimental Setting

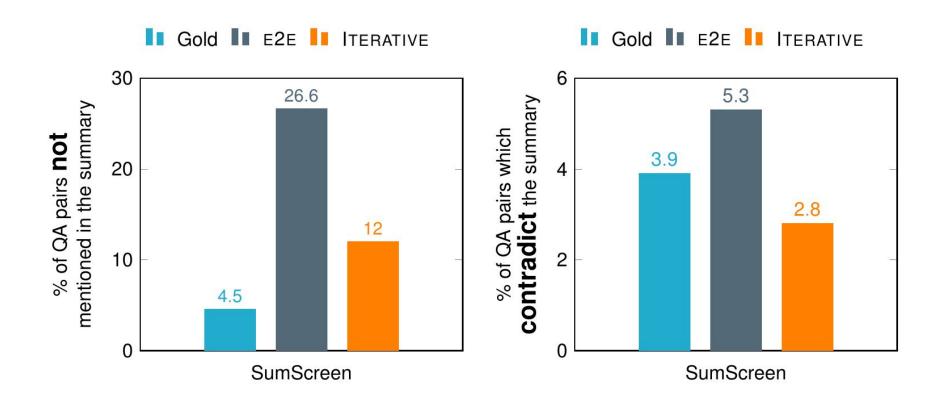
- AQuaMuse (Kulkarni et al., 2002, 2021): long-form question answering, simulates search engine, answer based on multiple retrieved documents.
- WikiCatSum (Perez-Beltrachini et al., 2019): topic-focused multi-document summarization, generate Wikipedia abstracts.
- SummScreen (Chen et al., 2022): dialogue summarization, generate summaries of TV episodes (e.g., CSI, The Bing Bang Theory).

All experiments with a LongT5 (Guo et al., 2021) backbone model.

Evaluation: How Good are the Blueprint Summaries?



Evaluation: Are Summaries Grounded to their Blueprint?



Interim Conclusions

- Bleuprint models are more informative than vanilla seq2seq models
- Errors can be examined and traced back to the blueprint
- QA pairs are intuitive and user-friendly, human-in-the-loop generation
- How do blueprints work across tasks and modalities?

Visual Story Telling: Blueprint Annotation











My Uncle Jack made us go to the Purdue game on Saturday. A₀: the Purdue game He went to Purdue and he thinks they are great no matter Q_1 : When did my uncle Jack made us go to the Purdue game? what else anyone thinks. We all had to sit in the bleachers A₁: Saturday and wear some ridiculous brown coats...

Q₀: Where did my uncle Jack made us go on Saturday?

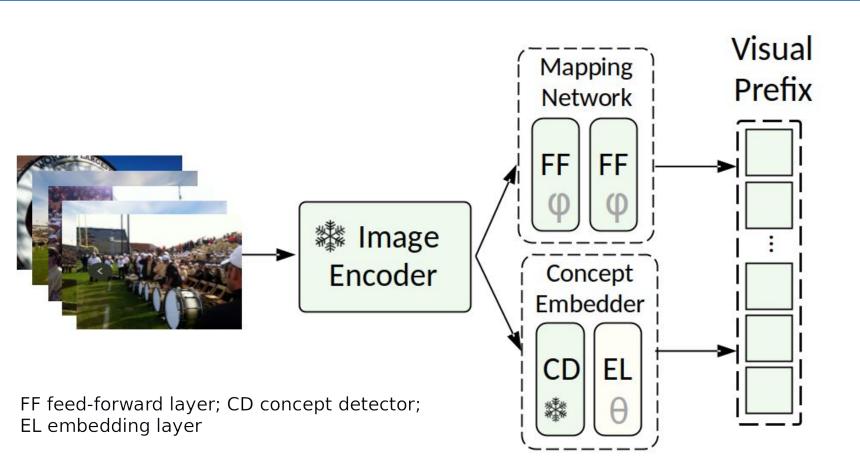
 Q_2 : Where did we all sit?

A₂: the bleachers

Q₃: What did the fans of Purdue have to wear?

A₃: some ridiculous brown coats

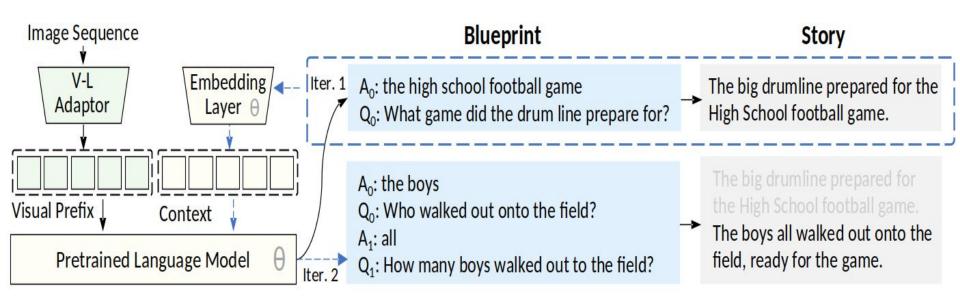
Visual Prefixes



Blueprint Storytelling Model (Iterative)

- Use a generation model (based on T5) to turn VIST stories into blueprints
- Train a linear model to map image sequences into visual prefixes (based on ResNet-152)
- Fine-tune a pretrained model to generate blueprints and stories from visual prefixes
- Generation proceeds sentence-by-sentence, where previous sentences are context for the current one: iterative model
- An end-to-end model that generates all sentences at once from the visual prefix performs worse

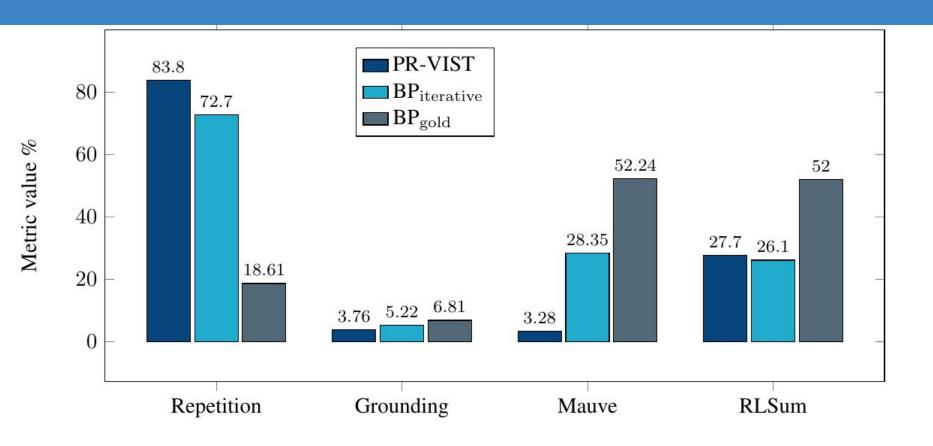
Iterative Blueprint Model



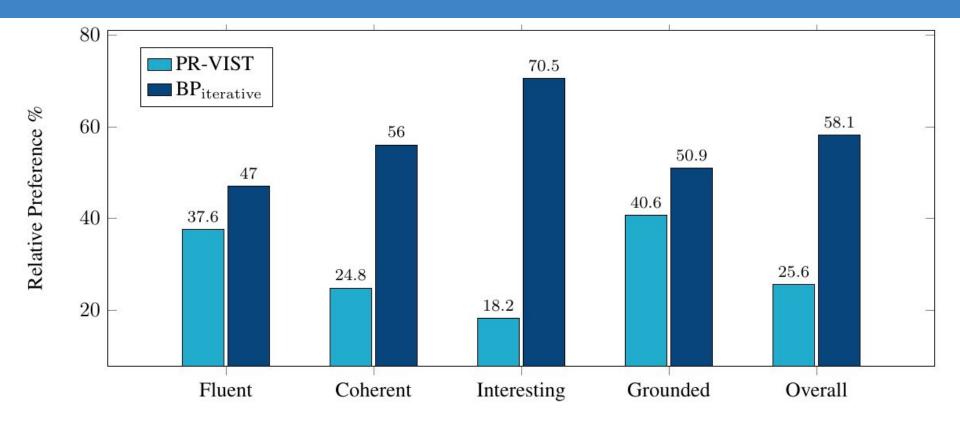
Evaluation

- Automatic evaluation using Blue, Rouge, Meteor is problematic mostly measures fluency, and all modern LMs generate fluent output
- Also, a story may be interesting, coherent, and grounded in the images, but completely different from the reference
- We use automatic evaluation using repetition and grounding, plus Mauve for naturalness (similarity of distribution with human text)
- More importantly, we ask human judges to evaluate fluency, coherence, interestingness and grounding
- We compare to a range of SotA models and to GPT 3.5 (with and without Blueprint)

Automatic Evaluation



Human Evaluation



Example Output











PR-VIST: I bought a cart to the market with some food. It looked great in the market. There were so many of their enthusiasm. They even had a sale with their bumper. The market was sold out of fresh vegetables. the [location] car is very nice and beautiful.

VP-BART: I went to the market yesterday. There were many different kinds of fruits there. I bought a lot of them. They were very expensive. Afterward I went back home.

Iterative: Today we decided to take a small shopping trip to the Market. The market had so many wonderful things to choose from that we looked at and bought so many of them. There was an array of different fruits that I could hardly resist buying. There was also a huge array of various types of pepper. Finally the day had come to an end and we piled in the taxi back to head home.

GPT-3.5 + BP: On Saturday morning, I visited the local market. It had a lot of fruits and vegetables being sold. The atmosphere was friendly and buzzing with many vendors ready to help you. The fresh produce was the best thing about the market. You can always find friendly vendors that sell the most delicious produce. The key to finding the best bargains at the market is bargaining. The market is always a great place to shop for fresh fruits and vegetables.

Human: Shoppers arrived early for the market. There was a variety of goods sold. Some carried away large bundles. Every item was fresh and colorful. The spot was a favorite among produce shoppers.

Analysis of Results

- Pretrained language models can produce better stories than specialized models trained from scratch
- The visual prefix is an effective interface between image and text; we don't need multimodal models
- Blueprint model output is most grounded, by automatic and human evaluation
- Iterative planning strategy that generates sentence by sentence works best
- GPT-3.5 struggles with Blueprints; they reduce its performance in most metrics

Conclusions

- Blueprint-based model generates output that is more coherent, interesting, and grounded than existing methods
- Blueprints aid in selecting key concepts and guiding narrative construction
- Blueprints are controllable: generate longer or shorter stories (more/less iterations), emphasize entities or characters (filter Blueprint), etc.
- Blueprints are interpretable and could enable human-in-the-loop and personalized storytelling

Narayan et al. Conditional Generation with a Question-Answering Blueprint. TACL,11, 2023. 974-996
Liu et al. Visual Storytelling with Question-Answer Plans. Findings of the Association for
Computational Linguistics: EMNLP 2023