# **Entity Skeletons for Visual Storytelling**

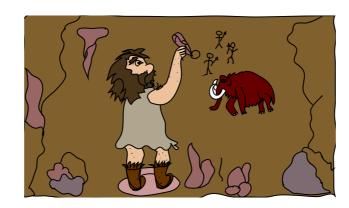
Ruo-Ping\*, Khyathi Chandu\*, Alan W Black

#### Overview

- Content as a Narrative Property
- Task Definition
- Dataset
- Models
  - Anchor Extraction
  - Anchor Informed Generation
- Results
  - Qualitative and Quantitative
  - Human Evaluation

# **History of Narratives**





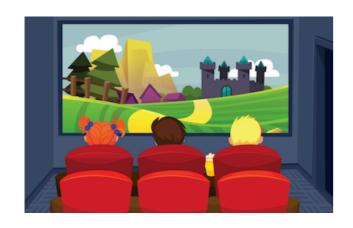


# **History of Narratives**













#### **Recent Advancements**

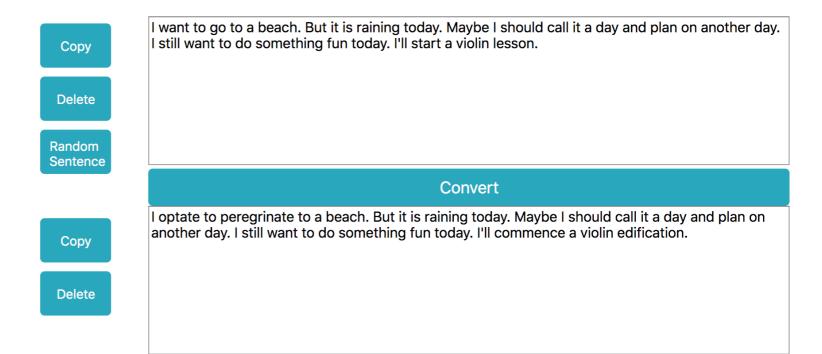
# Talk to Transformer See how a modern neural network completes your text. Type a custom snippet or try one of the examples. Learn more below. New from me: Create and deploy custom text classifiers for your app in minutes—no AI expertise needed! \$25 of free credits. Custom prompt We went to the beach. My kids had a lot of fun there.

#### Completion

We went to the beach. My kids had a lot of fun there. They loved the dunes and we loved the sand. Our backyard was full of dunes," Dr. Farina said.

GENERATE ANOTHER

Dr. Farina said the band wanted to move north from their original location in Orange County. But after hearing that Paradise Valley is a popular destination for musicians, they decided to build their new home on Route 22 near a closed golf



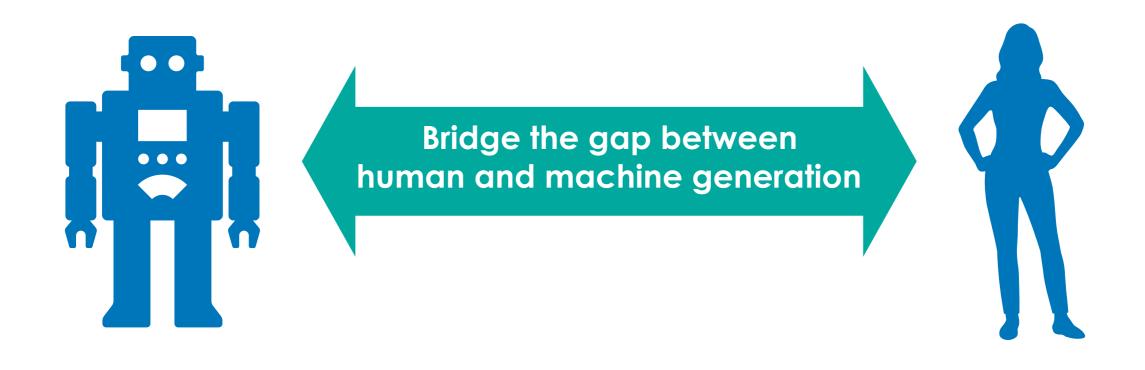
# Random Paragraph Generator Number of Paragraphs: 1 Generate Random Paragraphs Please LIKE & SHARE to keep our generators available! Click Like 61

There was something beautiful in his hate. It wasn't the hate itself as it was a disgusting display of racism and intolerance. It was what propelled the hate and the fact that although he had this hate, he didn't understand where it came from. It was at that moment that she realized that there was hope in changing him.

https://talktotransformer.com/

https://randomwordgenerator.com/paragraph.php

https://www.csgenerator.com/



#### What makes a narrative effective?

#### **Content - Relevance**

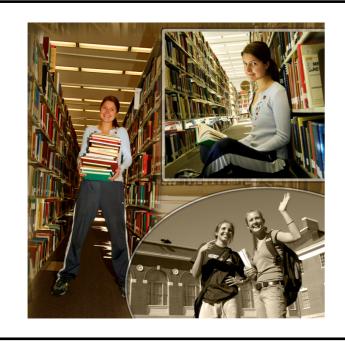


We went to the beach.

My kids had a lot of fun there.

There were a lot of palm trees.

We stayed in a resort.



We went to the library.
I love reading books.
I borrowed a lot of them.

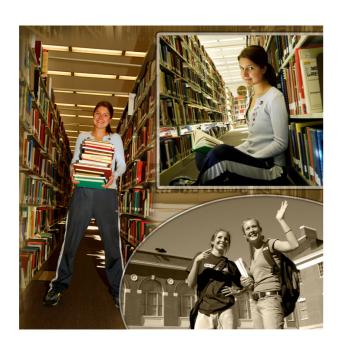
http://picturebite.com/Views-Sample-Albums

#### **Content - Relevance**



#### Entities

- Beach
- Kids
- Resort



#### Entities

- Library
- Student
- Books

# **Anchoring Framework**

#### **Fine-grained Entity Skeleton**

Input: 
$$I_i$$
 and  $E_i = \{e_i^{(1)}, e_i^{(2)}, ..., e_i^{(k)}\}$ 

Output: 
$$N_i = \{s_i^{(1)}, s_i^{(2)}, ..., s_i^{(k)}\}$$

Provides full guidance to each individual unit of narrative text

#### **Task Definition**

- Task: Introducing entities in visual stories
- Data:

$$S = \{S_1, ..., S_n\}$$

$$S_i = \{(I_i^{(1)}, x_i^{(1)}, y_i^{(1)}), ..., (I_i^{(5)}, x_i^{(5)}, y_i^{(5)})\}$$

- Input: Sequence of Images, Descriptions in Isolation (DII)
- Output: Stories in Sequences (SIS)
- Anchors: Entities

#### **Dataset**

- Visual Storytelling [1]
- Descriptions in Isolation (DII) absent for 25% of images

	Train	Val	Test	
# Stories	40,155	4,990	5,055	
#Images 200,775		24,950	25,275	
#without DII	40,876	4,973	5,195	

# **Entity Anchors**

- Entity Skeleton: defined as a linear chain of entities and referring expressions.
- Coreference chains are extracted from Stanford CoreNLP

# **Entity Anchors: 3 Forms**

Surface Form Coreference Chains

$$\{c_1, c_2, ..., c_5\}$$

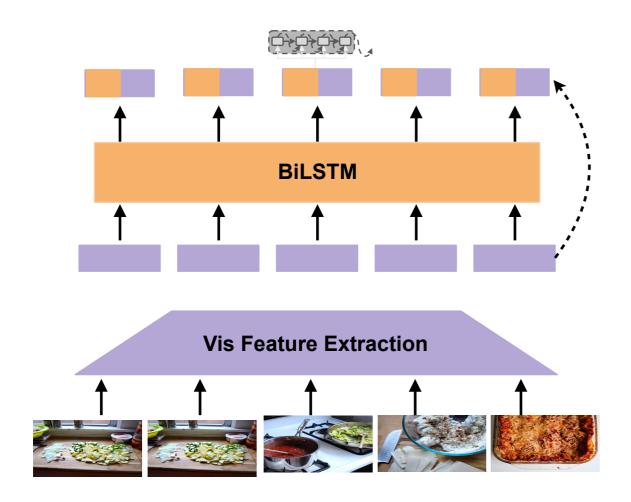
Nominalized Coreference Chains

$$\{[p,h]_1,...,[p,h]_5\}$$
  $p,h \in \{0,1\}$ 

Abstract Coreference Chains

{person, location, misc, object}

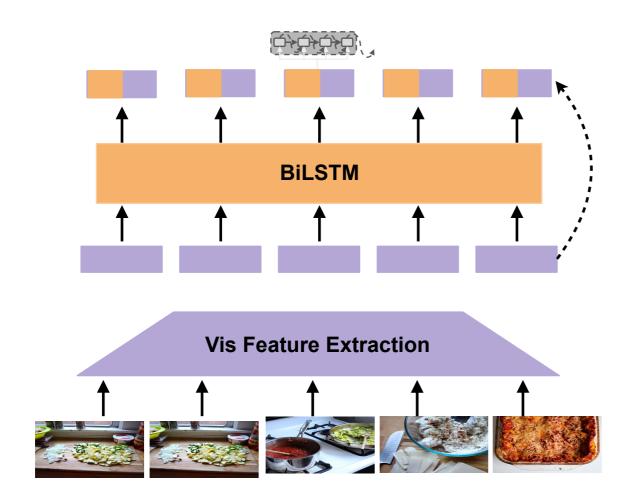
(1) Baseline: Glocal Context Model



$$egin{aligned} & oldsymbol{l}_t = \operatorname{ResNet}(oldsymbol{I}_t) \ & oldsymbol{g}_t = \operatorname{Bi-LSTM}([l_1, l_2 \dots l_5]_t) \end{aligned}$$

$$\hat{\boldsymbol{w}}_t \sim \prod_{\tau} Pr(\hat{\boldsymbol{w}}_t^{\tau} | \hat{\boldsymbol{w}}_t^{<\tau}, \boldsymbol{l}_t, \boldsymbol{g}_t)$$

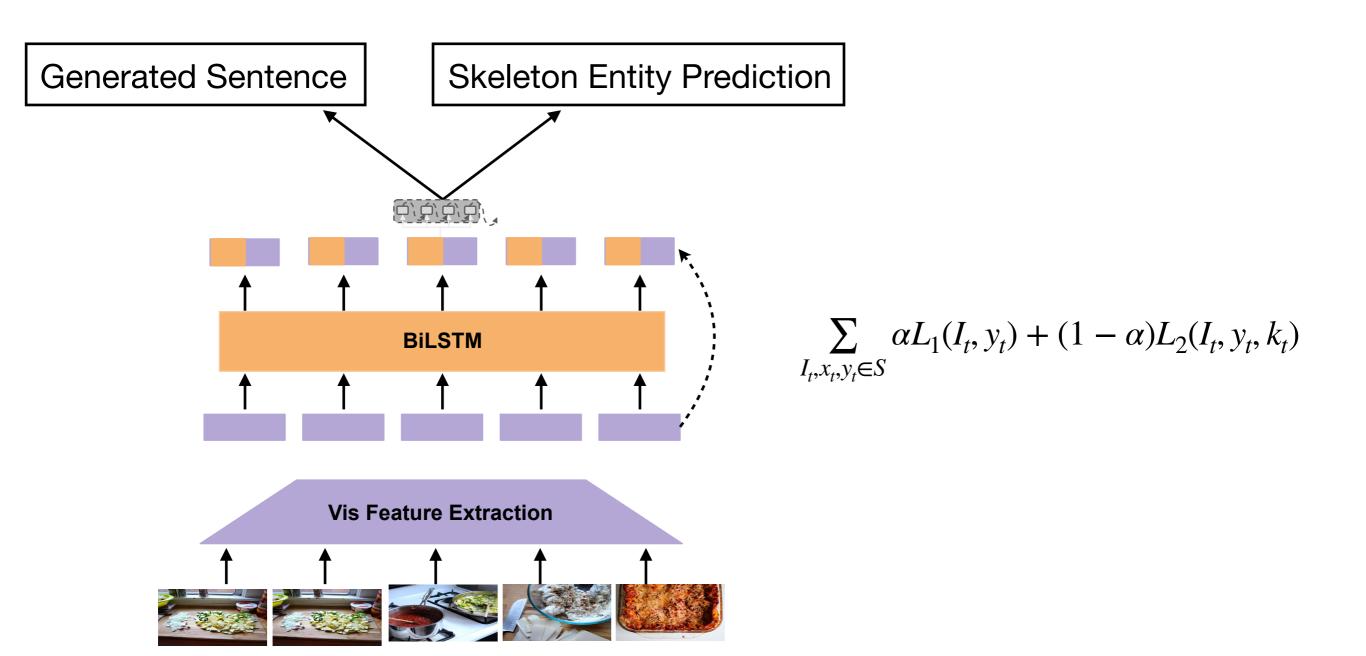
(2) Baseline: Skeleton Informed Local Context Model



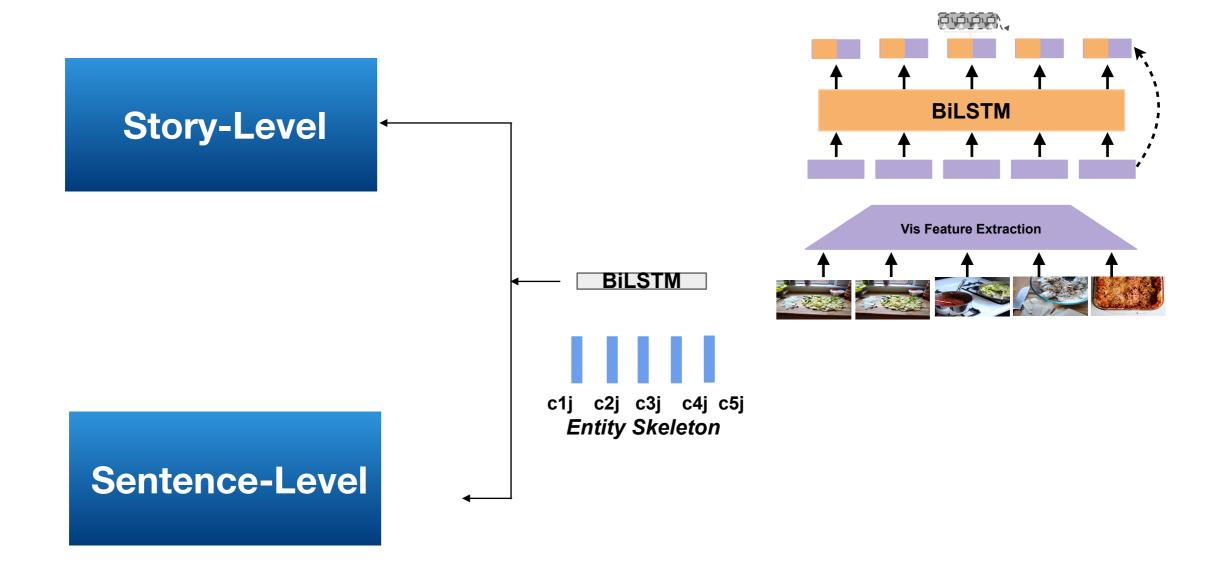
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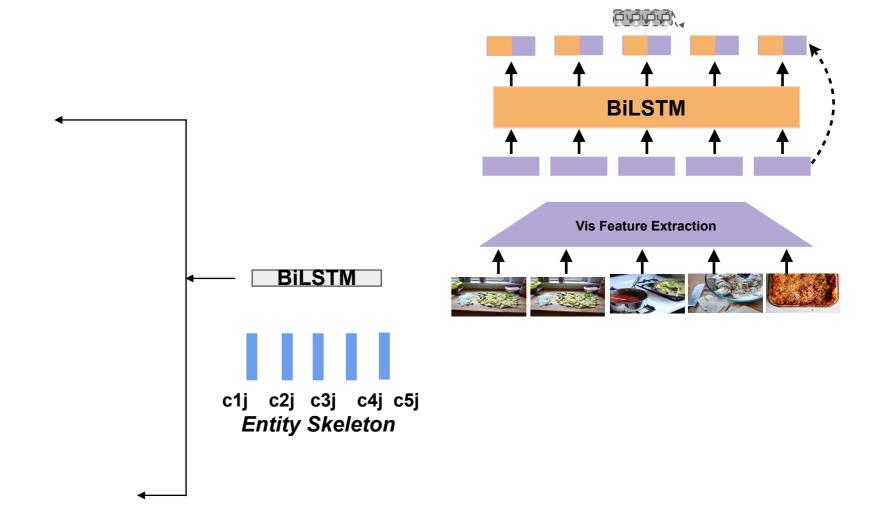
(3) Multitasking Skeleton Prediction



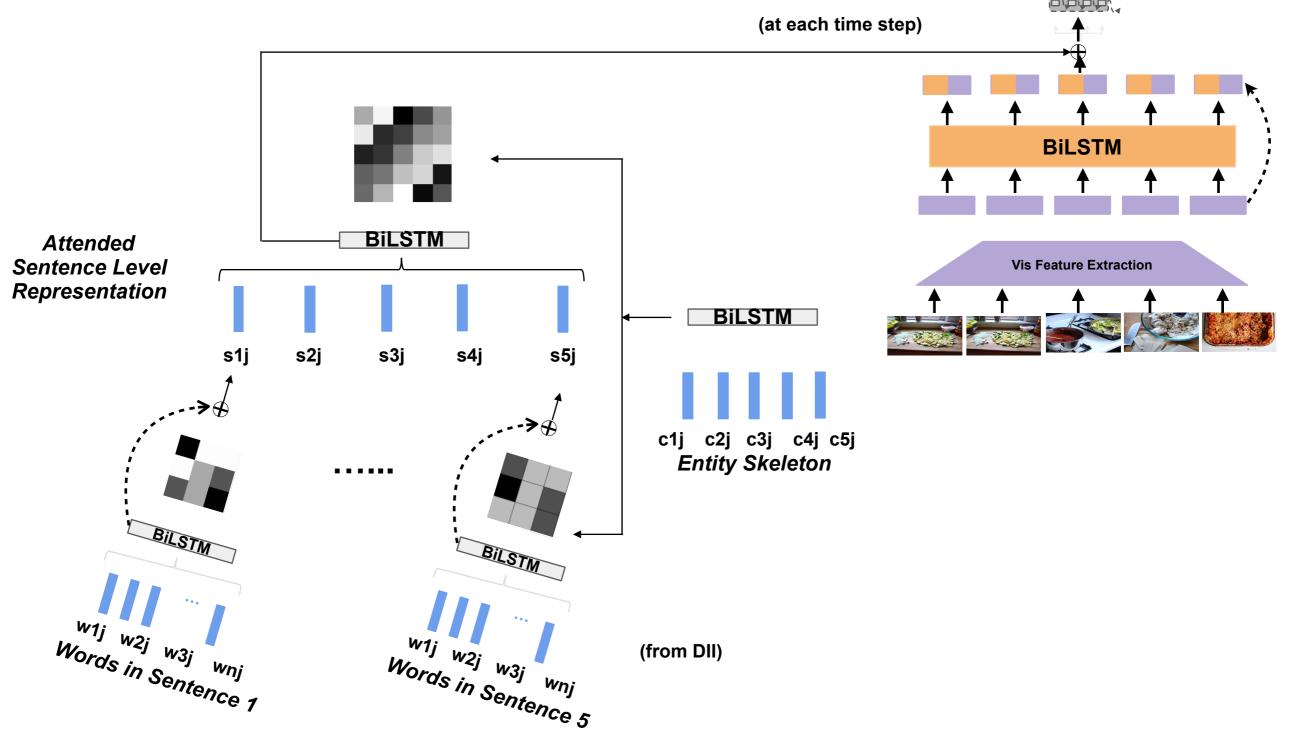
(4) Hierarchical Glocal Model



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(4) Hierarchical Glocal Model



#### **Quantitative Experimental Results**

Models	Skeleton Form	METEOR	Distance	Avg # of Distinct Entities
Baseline	None	27.93	1.02	0.4971
Baseline + Skeleton	Surface	27.66	1.02	0.5014
MTG ( $\alpha$ =0.5)	Surface	27.44	1.02	0.9554
MTG (α=0.4)	Surface	27.59	1.02	1.1013
MTG ( $\alpha$ =0.2)	Surface	27.54	1.01	0.9989
MTG (α=0.5)	Nominalization	30.52	1.12	0.5545
MTG ( $\alpha$ =0.5)	Abstract	27.67	1.01	0.5115
Glocal Attention	Surface	28.93	1.01	0.8963

Ground Truth: 0.7944

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#### **Human Evaluation**

- Preference Testing for Hierarchical Glocal Model
  - 82% over Baseline
  - 64% over Multitasking Model

#### **Qualitative Analysis**

Models					Minute Maid  THE PLAY IS URGAN OFFICIAL PRIMARY  THE PLAY IS URGAN	Phenomena
SIS		incredible . you could	_	action .	once the national anthem was sung , and the first pitch was thrown , the excitement began . it was a great game !	
Baseline Model	the city was a great place to visit .	i had a great time .	there were many people there .	we got to see a lot of cool things.	it was a lot of fun .	- Characters in the story are mentioned as "many people" instead of "we" (sentence 3).
Glocal Hierarchical Attention Model	we saw the building was packed.	i was excited to see my favorite team .	we were all excited to see the game.	we all got together to watch.	it was a great game .	<ul> <li>+ characters ('we' and 'it') were introduced at the right time</li> <li>+ Important entities were mentioned (building, game)</li> </ul>

#### **Qualitative Analysis**

Models		THE PARTY OF THE P			Minute Maid  The Pay is unoun  OPPICIAL RAUSE  1121456789 1-1-1-1-1-1-1-1-1-1-1-1-1-1-1-1-1-1-1-	Phenomena
SIS		incredible . you could			once the national anthem was sung , and the first pitch was thrown , the excitement began . it was a great game !	
Baseline Model	the city was a great place to visit .	i had a great time .	there were many people there .	we got to see a lot of cool things.	it was a lot of fun .	- Characters in the story are mentioned as "many people" instead of "we" (sentence 3).
Glocal Hierarchical Attention Model	we saw the building was packed.		we were all excited to see the game.	we all got together to watch.	it was a great game .	<ul> <li>+ characters ('we' and 'it') were introduced at the right time</li> <li>+ Important entities were mentioned (building, game)</li> </ul>

# **Takeaways**

- Improves Relevance component of visual storytelling
- Improves Controllability in generation
- Step towards interpretability with respect to intermediate representation

### **Thank You**

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