

# Creative and Artistic Text Generation

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# What Contents Are Included in This Tutorial?

- Poetry Generation
- Story Generation
- Multi-Modal Generation
   Visual Storytelling
   Visual Poetry Generation
- Other Genres
  - Couplet
  - Lyrics

# Target Audience

- Ph.D. students or researchers who are working on artistic text generation.
- Anyone who wants to learn how neural approaches (i.e., deep learning techniques) can be applied to artistic text generation.
- Anyone who wants to build an artistic text generation system (e.g., story, poetry, couplet) with state-of-the-art neural techniques.

# Outline

- Introduction
- Background Knowledge
- Existing Methods

   Poetry Generation
   Story Generation
   Multi-Modal Generation
   Other Genres
- Recent Trends and Conclusion
- Q&A

# Roadmap

- Introduction
- Background Knowledge
- Existing Methods
   Poetry Generation
   Story Generation
   Multi-Modal Generation
   Other Genres
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#### Task Description of Poetry Generation



- Input: a piece of text
- Output: a poem that meets given constraints

Marjan Ghazvininejad et al. ACL' 17 Xingxing Zhang et al. EMNLP' 14

#### Example System: Jiuge



https://jiuge.thunlp.cn/jueju\_en.html

# Task Description of Story Generation

Definition 1: [Martin et al. AAAI' 18]
 Automated story generation is the problem of automatically selecting a sequence of events, actions, or words that can be told as a story.

• Definition 2: [Xu et al. EMNLP' 18]

**Input**: A short description of a scene or an event.

**Output**: A relevant narrative story following the input.

#### Examples

**Input**: Fans came together to celebrate the opening of a new studio for an artist.

**Output**: The artist provided champagne in flutes for everyone. Friends toasted and cheered the artist as she opened her new studio.

**Input**: Last week I attended a wedding for the first time.

**Output**: There were a lot of families there. They were all taking pictures together. Everyone was very happy. The bride and groom got to ride in a limo that they rented.

Table 1: An illustration of narrative story generation.

#### **Demo Story Generation System**

	Stories v1.0 Auto Interactive Advanced -			4.82 seco
Input	summer	Generate		
	Ready			
	Storyline school -> was -> asked -> said -> told			
Output	Title to Story it was a hot summer day . the kids were bored . they decided to go to the park . they played in the water . they had a great time .	Plan and Write i went to school the other day. it was my first day of school. i asked my mom if she could go with me. she said yes. she said yes.	Plan and Revise i went to school the other day. it was my first day of high school. i asked my mom to take me. she said no.	

Disclaimer: All the entries made in this website will be logged for academic research.

<u>http://cwc-story.isi.edu/</u> Goldfarb-Tarrant et al., *NAACL-HLT ' 2019* 

#### Image-Inspired Poetry Generation



# Visual Storytelling

Input

![](_page_10_Picture_2.jpeg)

![](_page_10_Picture_3.jpeg)

Xin Wang et al., ACL' 18 Bairui Wang et al., AAAI' 19

#### Others

![](_page_11_Picture_1.jpeg)

![](_page_11_Picture_2.jpeg)

And humble and their fit *flees* are wits size but that one made and made thy step me lies

Cool light the golden dark in any way the birds a *shade* a laughter turn away

Then adding wastes retreating white as thine She watched what eyes are breathing awe what shine

But sometimes shines so covered how the beak Alone in pleasant skies no more to seek

![](_page_11_Picture_7.jpeg)

![](_page_11_Picture_8.jpeg)

Verse

#### Others

But she fell in love with him Girl when they feel the same The princess was in love with the priest Can't let go and it never goes out

She also abominated what he did Be the things they said The princess was shocked by the priest's actions And though her heart cant take it all happens<sup>1</sup>

![](_page_12_Picture_3.jpeg)

![](_page_12_Figure_4.jpeg)

pessoa libertista em dó menor reparte a pessoa e o justador instrumentos de tenta para o refrão é preciso danças pipocas perdendo-se as esferas da transmissão o que foi papa de xaréns libertarian person in C minor redistributes the person and the fighter instrument of probe to the chorus it takes dances popcorn if lost the transmission domains what was xarém food

Song Lyrics

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#### Sequence to Sequence Model

- Common Used Sequence Generation Method
- Stable and Easy for Training
- Flexibility

![](_page_15_Figure_4.jpeg)

Sutskever, Ilya et al. NIPS, 2014 Bahdanau, Dzmitry et al., ICLR, 2015

#### **Convolutional Sequence to Sequence**

- Fast Training
- Strong Language Model for Capturing Long-Range Dependencies
- Bounded CNN Context Window

![](_page_16_Figure_4.jpeg)

Gehring, Jonas, et al., ICML, 2017

# Transformer

- Fast Training
- Strong Language Model for Capturing Long-Range Dependencies
- Correlations Learning
- The SOTA Language Model

![](_page_17_Figure_5.jpeg)

![](_page_17_Figure_6.jpeg)

#### Variational Autoencoder

- Generative Model
- Wording Diversity
- Intra-Sentence Consistency
- Address Sparsity

![](_page_18_Figure_5.jpeg)

#### Generative Adversarial Nets

- One-to-Many Generation
- Enhancing Generator
- Supervision Signal

![](_page_19_Figure_4.jpeg)

Yu, Lantao, et al., AAAI, 2017

# **Reinforcement Learning**

- Directly Model Discrete Sequence
- Address Loss-Evaluation Mismatch

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#### **Recurrent Neural Model**

Task Chinese Quatrain

Generation Process
 Keywords
 Keywords expansion
 Incremental generation

![](_page_23_Figure_3.jpeg)

![](_page_23_Figure_4.jpeg)

Xingxing Zhang et al. EMNLP' 14

#### **Recurrent Neural Model**

- Convolutional Sentence Model (CSM)  $v_i = \text{CSM}(S_i)$
- Recurrent Context Model (RCM)  $u_i^j = \text{RCM}(v_{1:i}, j)$
- Recurrent Generation Model (RGM)  $P(w_{j+1}|w_{1:j}, S_{1:i}) = \text{RGM}(w_{1:j+1}, u_i^{1:j})$
- Training

Cross Entropy Errors

![](_page_24_Figure_6.jpeg)

# Planning-Based Recurrent Neural Model

![](_page_25_Figure_1.jpeg)

- Keyword Extraction: TextRank Algorithm
- Keyword Expansion: RNNLM-Based Method; Knowledge-Based Method
- Poetry generation: Bidirectional RNN (GRU) Encoder; Attention; RNN (GRU) Decoder

Zhe Wang et al. COLING' 16.

# **Iterative Polishing**

 Intention Representation CNN RNN

 Sequential Generation Hierarchical RNN Character by Character

• Iterative Polishing Re-Generation

![](_page_26_Figure_4.jpeg)

Rui Yan et al. IJCAl' 16.

#### Interactive Poetry Generation

- Step 1
   Search related rhyme words
- Step 2 Create a finite-state acceptor (FSA)
- Step 3 RNN guided by FSA

![](_page_27_Figure_4.jpeg)

#### GAN for Poetry Generation

- GAN
   Min-Max Game
- Generator Reinforcement learning MC search
- Discriminator

![](_page_28_Figure_4.jpeg)

Lantao Yu et al, AAAl' 17

# Static Memory Model

![](_page_29_Figure_1.jpeg)

# Working Memory Model

![](_page_30_Figure_1.jpeg)

Xiaoyuan Yi et al., IJCAI' 18

#### **Conditional Variational Autoencoder**

![](_page_31_Figure_1.jpeg)

#### **CVAE-GAN Model**

![](_page_32_Figure_1.jpeg)

# Mutual Reinforcement Learning

- Modeling Poetry Generation as RL Problem
- Fine-Grained Reward Designing
   Fluency Rewarder (LM)
   Coherence Rewarder (MI)
   Meaningfulness Rewarder (TF-IDF)
   Overall Quality Rewarder (Classifier)

 Mutual Reinforcement Learning Two Generators Instance-Based Method Distribution Level Mutual learning

![](_page_33_Figure_4.jpeg)

Xiaoyuan Yi et al., EMNLP' 18

## **Stylistic Poetry Generation**

![](_page_34_Figure_1.jpeg)

#### Pretraining-Based Model

- Pre-trained Model GPT
- Genres
   Quatrain
   lambics
   Couplet
- Fine-Tuning Model Transformer Auto-regressive Language Model

![](_page_35_Figure_4.jpeg)

Yi Liao et al., arXiv:1907.00151

# **Rhetorically Controlled Generation**

- Modern Poetry Generation
- Manual Control CVAE Model
   Process User Input As Rhetorical Label
- Automatic Control CVAE
   Predict *When* Use Rhetoric Label
- Topic Memory
   Store Topic Information
- Rhetorically Controlled Decoder
   Generate Sentence with Forms of Rhetoric

![](_page_36_Figure_6.jpeg)

Zhiqiang Liu et al., ACL' 19

# Human-Machine Collaborative Generation

![](_page_37_Figure_1.jpeg)

Zhipeng Guo et al., ACL' 19

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   Multi Model Generation

Other Genres

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#### **Coherent Story Generation**

![](_page_41_Figure_1.jpeg)

![](_page_41_Figure_2.jpeg)

![](_page_41_Picture_3.jpeg)

Jain, Parag, et al., arXiv:1707.05501, 2017

#### **Event Representations**

- Story to Event Sequences
   5-tuple Event Representations
- Event to Event Generation
   Event-Level Seq2seq
- Event to Story Generation Seq2seq

![](_page_42_Figure_4.jpeg)

# **Controllable Story Generation**

- Input Human inputs Controllable factors
- Output

٠

A story that coherent to human inputs

- Ending Valence Control Data labeling Supervised classifier Conditional LM for generation
- Storyline Control Keywords extractor Conditional LM for generation

![](_page_43_Figure_6.jpeg)

Peng, Nanyun, et al., Workshop, 2018

# **Hierarchical Story Generation**

- Hierarchical Generation Pipeline
   Generating Prompts---Story
- Convolutional Seq2seq For Generating Prompts
   Conventional Convolutional Seq2seq Model
- Gated Multi-Scale Attention
   Gated self-attention to attend Information at different position
   Multi-scale attention to attend information at different granularity
- Prompts Fusion

Residual Learning Upon pre-trained Convolutional seq2seq model

**Prompt:** The Mage, the Warrior, and the Priest

**Story:** A light breeze swept the ground, and carried with it still the distant scents of dust and time-worn stone. The Warrior led the way, heaving her mass of armour and muscle over the uneven terrain. She soon crested the last of the low embankments, which still bore the unmistakable fingerprints of haste and fear. She lifted herself up onto the top the rise, and looked out at the scene before her. [...]

Fan Angela et al., ACL, 2018

# Skeleton to Story Generation

- Skelton-Based Generative Module Input-to-Skeleton Skeleton-to-Sentence
- Skeleton Extraction Module
   Pretraining on Sentence Compression Dataset
   Reinforcement Learning Training
   Two Entropy Loss Reward
   Iterative Optimization

![](_page_45_Figure_3.jpeg)

Xu, Jingjing, et al., EMNLP, 2018

#### Planning-Based Method

![](_page_46_Figure_1.jpeg)

- Plan and Write •
- Static Planning ٠
- **Dynamic Planning** ٠

(b) Static schema work-flow.

**S**3

*s*<sub>2</sub>

 $S_1$ 

 $l_5$ 

**S**5

 $l_5$ 

**S**5

l₄

*s*<sub>4</sub>

# **CVAE and Memory Network**

![](_page_47_Figure_1.jpeg)

#### Plan Write and Revise

- System Combination
- Cross-Model Mode
- Intra-Model Mode
- Story Writer
   Title-to-Story
   Plan-and-Write
   Plan-and-Revise

![](_page_48_Figure_5.jpeg)

Goldfarb-Tarrant, Seraphina et al., NAACL-HLT, 2019

# **BERT Augmented Story Ending Prediction**

![](_page_49_Figure_1.jpeg)

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## Image Inspired Poetry Generation

- Image-Based Encoder CNN Bidirectional RNN
- Memory-Based Decoder Keyword Extractor Vector Representations

![](_page_52_Figure_3.jpeg)

#### Visual Poetry Generation of Xiaolce

![](_page_53_Figure_1.jpeg)

Framework

Cheng, Wen-Feng, et al., arXiv:1808.03090, 2018

#### Multi-Modal Poetry Generation

![](_page_54_Figure_1.jpeg)

Liu, Dayiheng, et al., IJCNN, 2018

# Multi-Adversarial Training

![](_page_55_Figure_1.jpeg)

Liu, Bei, et al., ACM, MM, 2018

#### Inverse Reinforcement Learning

![](_page_56_Figure_1.jpeg)

Overall Framework

![](_page_56_Figure_3.jpeg)

Policy Model

Wang, Xin, et al., ACL, 2018

#### Hierarchical Photo-Scene Encoder

![](_page_57_Figure_1.jpeg)

Wang, Bairui, et al., AAAI, 2019

## Hierarchically Structured Reinforcement Learning

![](_page_58_Figure_1.jpeg)

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- Introduction
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![](_page_61_Picture_0.jpeg)

Tasks	Main Techniques	
Rap Lyric Generation [Potash Peter et al.,15]	LSTM + Explicit Templates	
Rap Lyric Generation [Malmi Eric et al., 16]	Information Retrieval Task	
Chinese Song lambics Generation [Wang et al.,16]	Attention-Based Seq2seq	
Chinese Couplet Generation [Yan Rui et al., 16]	Seq2seq + Attention + Polishing	
Rhythmic Verse Generation [Hopkins Jack, 17]	Multi-LSTM LM + Finite State Transducers	
Theme-Aware Lyrics Generation [Wang Jie, 19]	Multi-Channel Seq2seq + LDA	

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# Recent Trends

Learning Method

 Reinforcement Learning
 Inverse Reinforcement Learning
 Mutual Learning
 Imitation Learning

- Fine-Grained Controlling Sentiment Ending Forms
- Pre-Training Method GPT BERT XLNet

#### Recent Trends

- Better Evaluation Metrics
- Commonsense and Knowledge
- Reasoning
- New Forms of Artistic Text
- Explainability

#### Conclusion

- Poetry Generation
- Story Generation
- Multi-Modal Generation Image-Inspired Poetry Generation Visual Storytelling
- Other Genres

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# Thank you!