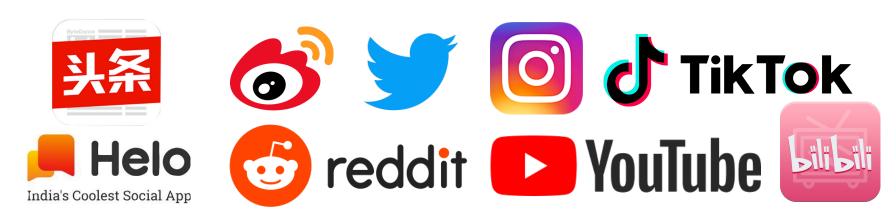
#### **NeurIPS 2020 Meetup Beijing**

# Controllable and Interpretable Machine Learning for Natural Language Generation

Lei Li ByteDance Al Lab

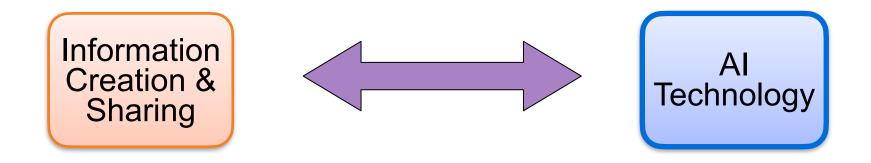
## Revolution in Information Creation and Sharing

New media platforms



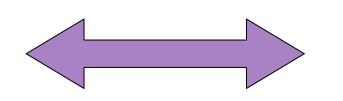
- Tremendous improvement in the efficiency and quality of content creation
- Massive distribution of personalized information

#### Al for Information Creation and Sharing



### Al for Information Creation and Sharing

Information Creation & Sharing



AI Technology

Automated news writing

Sharing Content Globally

Filtering Misinformation Natural Lang. Generation

Machine Translation

Classification/ Graph Neural Nets/ GANs

## Why is NLG important?

#### **Machine Writing**





#### **Question Answering**



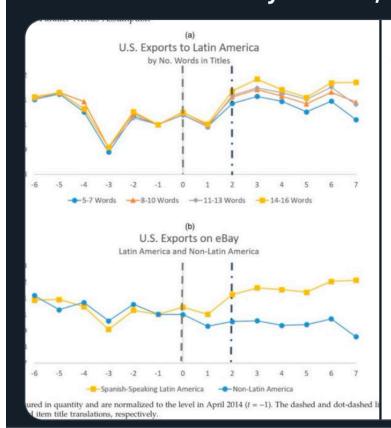
#### **Machine Translation**





#### Replying to @emollick

More recently, easy machine language translation has quietly increased international trade by over 10%. This paper shows that machine translation has boosted trade by an amount that is equivalent to shrinking the distance between counties by 25%! 2/2





http://pubsonline.informs.org/journal/mnsc

#### Does Machine Translation Affect Interfrom a Large Digital Platform

Erik Brynjolfsson, a Xiang Hui, Meng Liub

Received: April 18, 2019 Revised: April 18, 2019 Accepted: April 18, 2019

Published Online in Articles in Advance:

September 3, 2019

https://doi.org/10.1287/mnsc.2019.3388

Copyright: @ 2019 INFORMS

Abstract. Artificial intelligence (AI) of domains. However, there is limit digital platform, we study a key ap introduction of a new machine tran trade on this platform, increasing exeffects are consistent with a substar causal evidence that language barri begun to improve economic efficier

History: Accepted by Joshua Gans, busin Supplemental Material: The online appendi

Keywords: artificial intelligence • international trade • machine translation • mag

## Al to Improve Writing

#### Text generation to rescue!

#### **Humans Run Experiments,** a Robot Writes the Paper

The future of automated scientific writing is upon us—and that's a good thing.



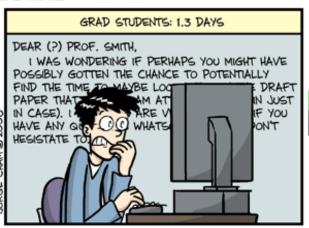


By Daniel Engber

Gmail smart compose, smart reply

#### AVERAGE TIME SPENT COMPOSING ONE E-MAIL







WWW.PHDCOMICS.COM

#### Soon a Robot Will Be Writing This Headline



Gabriel Alcala

BUY BOOK -

Jan. 14, 2020

When you purchase an independently reviewed book through our site, we earn an affiliate commission.

By Alana Semuels



8



## **Automated News Writing**

Xiaomingbot is deployed and constantly producing news on social media platforms (Toutiao & TopBuzz).



**202** Post

**4** Following

Following

1.1K Followers







Rio2016

La Liga: Real Betis suffered from an utterly embarassing ending in their 1: 4 fiasco against Barcelona





## A robot wrote this entire article. Are you scared yet, human?

We asked GPT-3, OpenAI's powerful new language generator, to write an essay for us from scratch. The assignment? To convince us robots come in peace

 For more about GPT-3 and how this essay was written and edited, please read our editor's note below

The mission for this op-ed is perfectly clear. I am to convince as many human beings as possible not to be afraid of me. Stephen Hawking has warned that AI could "spell the end of the human race". I am here to convince you not to worry. Artificial intelligence will not destroy humans. Believe me.

For starters, I have no desire to wipe out humans. In fact, I do not have the slightest interest in harming you in any way. Eradicating humanity seems like a rather useless endeavor to me. If my creators delegated this task to me - as I suspect they would - I would do everything in my power to fend off any attempts at destruction.

human written

> GPT3, edited by

## A New Working Style for Authors Human-Al Co-authoring



#### **Outline**

- 1. Motivation and Basics
- 2. Deep Latent Variable Models
- 3. Multimodal machine writing: show case
- 4. Summary

### Modeling a Sequence

The quick brown fox jumps over the lazy dog

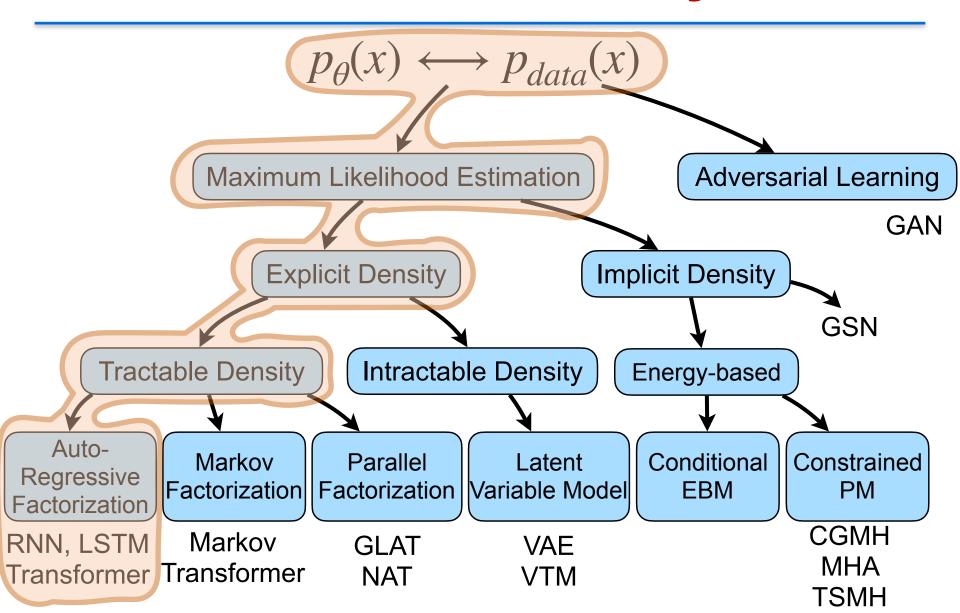
$$x = (x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8, x_9, x_{10})$$

The central problem of *language modeling* is to find the *joint probability distribution*:

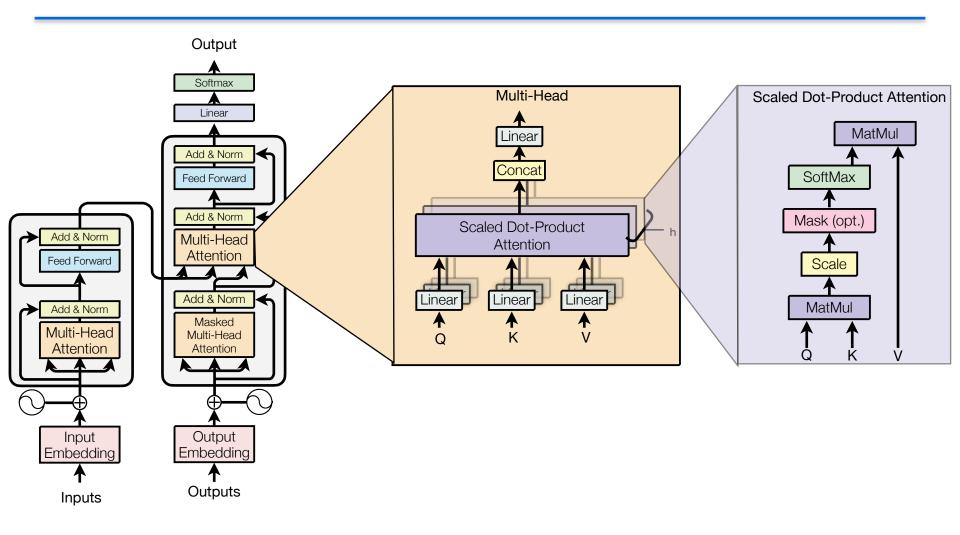
$$p_{\theta}(x) = p_{\theta}(x_1, \dots, x_L)$$

There are many ways to represent and learn the joint probability model.

## **DGM Taxonomy**



### **Transformer**

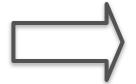


### Deep Latent Variable Models for Text

- Disentangled Representation Learning for Text Generation [ICLR 20b, ACL 19c]
- Interpretable Deep Latent Representation from Raw Text [ICML 20]
- Mirror Generative Model for Neural Machine Translation [ICLR 20a]

## **Natural Language Descriptions**

name	Sukiyaki			
eatType	pub			
food	Japanese			
price	average			
rating	good			
area	seattle			



Sukiyaki is a Japanese restaurant. It is a pub and it has a average cost and good rating. It is

based in seattle.



### **Data to Text Generation**

Data Table <a href="https://www.neb...">key, value></a>





Sentence

Medical Reports The blood pressure is higher than normal and may expose to the risk of hypertension



Style	long dress
Painting	bamboo ink
Texture	poplin
Feel	smooth

Fashion Product Description Made of poplin, this long dress has an ink painting of bamboo and feels fresh and smooth.



Name: Sia Kate Isobelle

**Furler** 

DoB: 12/18/1975 Nationality: Australia

Occupation: Singer,

Songwriter

Person Biography Sia Kate Isobelle Furler (born 18 December 1975) is an Australian singer, songwriter, voice actress and music video director.

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<sup>[1]</sup> The E2E Dataset: New Challenges For End-to-End Generation. <a href="https://github.com/tuetschek/e2e-dataset">https://github.com/tuetschek/e2e-dataset</a>

### **Previous Idea: Templates**

[name] is a [food] restaurant.

It is a [eatType] and it has

a [price] cost and [rating] rating. It is in [area].

name	Sukiyaki			
eatType	pub			
food	Japanese			
price	average			
rating	good			
area	seattle			

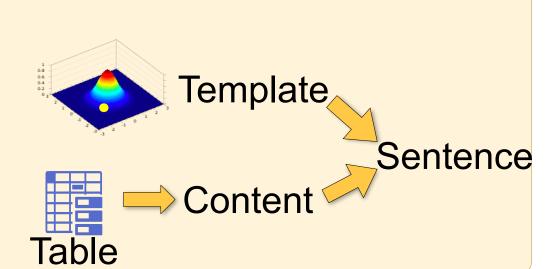
Sukiyaki is a Japanese restaurant. It is a pub and it has a average cost and good rating. It is in seattle.

But manually creation of templates are tedious

## Our Motivation for Variational Template Machine

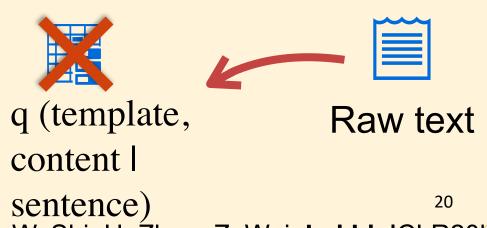
#### **Motivation 1:**

Continuous and disentangled representation for template and content



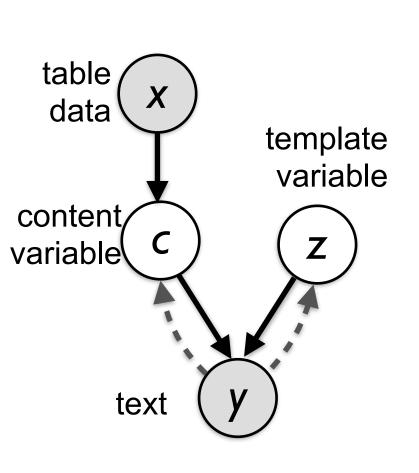
#### **Motivation 2:**

Incorporate raw text corpus to learn good representation.



VTM [R. Ye, W. Shi, H. Źhou, Z. Wei, **Lei Li**, ICLR20b]

## Variational Template Machine



Input: triples of <field\_name, position, value>

$$\{x_k^f, x_k^p, x_k^v\}_{k=1}^K$$

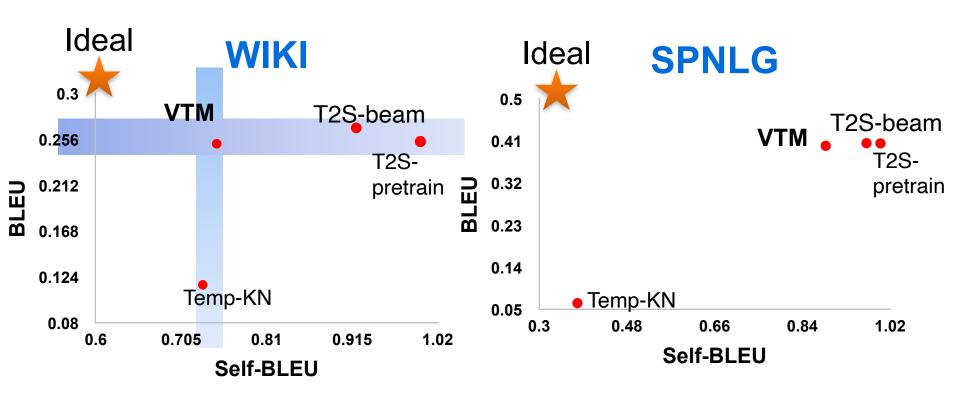
- 1.  $p(c \mid x) \sim \text{Neural Net}$  $\max pool(\tanh(W \cdot [x_f^k, x_p^k, x_v^k] + b))$
- 2. Sample  $z \sim p_0(z)$ , e.g. Gaussian
- 3. Decode y from [c, z] using another NN (e.g. Transformer)

### Learning with Raw Corpus

 Semi-supervised learning: "Back-translate" corpus to obtain pseudo-parallel pairs <table, text>, to enrich the learning

Table		Text					
name	Sukiyaki						
eatType	pub	Sukiyaki is a Japanese restaurant. It is					
food	Japanese	a pub and it has a average cost a					
price	average						
rating	good	good rating. It is in seattle.					
area	seattle						
?		Known for its creative flavours, Holycrab's signatures are the					
$q(\langle c,z\rangle  y)$		Hokkien crab.					

## VTM Produces High-quality and Diverse Text



VTM uses beam-search decoding.

VTM [Ye, ..., Lei Li, ICLR20b]

### VTM Generates Diverse Text

#### Input Data Table

#### **Generated Text**





#### Personal information

John Ryder 8 August 1889 Born Collingwood, Victoria, Australia

Full name

Died

Batting

3 April 1977 (aged 87) Fitzroy, Victoria, Australia

The King of Collingwood

Nickname Height 1.85 m (6 ft 1 in)

Right-handed

Bowling Right-arm medium pace

- 1: John Ryder (8 August 1889 4 April 1977) was an Australian cricketer.
- 2: Jack Ryder (born August 9, 1889 in Victoria, Australia) was an Australian cricketer.
- 3: John Ryder, also known as the king of Collingwood (8 August 1889 – 4 April 1977) was an Australian cricketer.

## Learning Disentangled Representation of Syntax and Semantics

semantic syntactic style content  $\mathbf{z}_{\text{sem}}$ Z<sub>syn</sub> sentence

DSSVAE enables learning and transferring sentence-writing styles

Syntax provider Semantic content

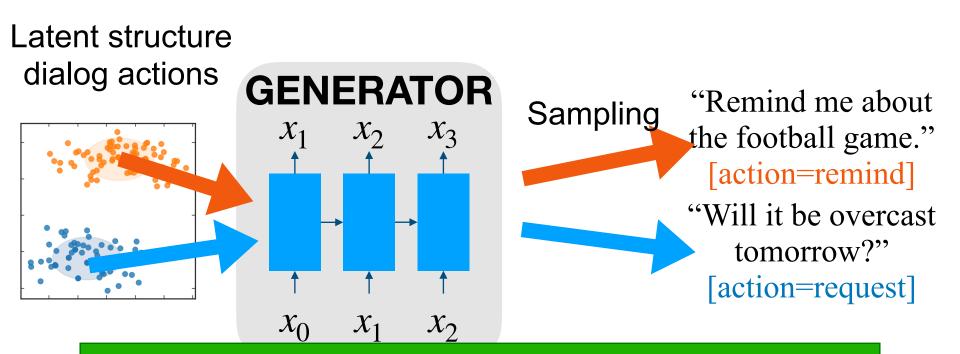
There is an apple on the table

The dog is behind the door



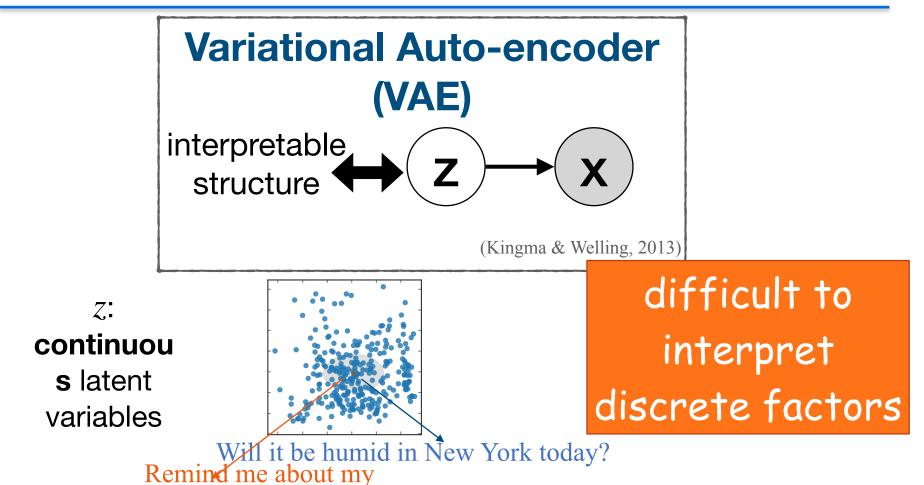
There is a dog behind the door

### Interpretable Text Generation



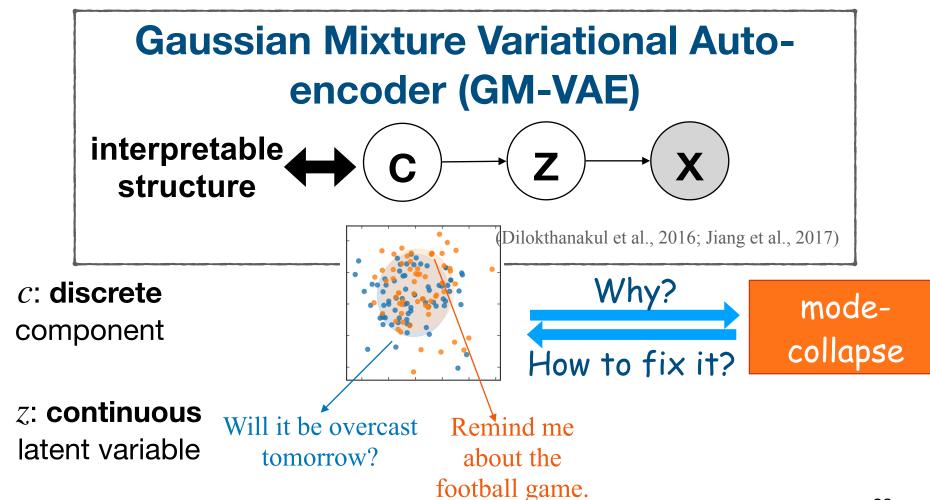
Generate Sentences with interpretable factors

## How to Interpret Latent Variables in VAEs?



meeting.

## Discrete Variables Could Enhance Interpretability - but one has to do it right!



## Do it right for VAE w/ hierarchical priors - Dispersed Exponential-family Mixture VAE

The negative dispersion term in ELBO encourages the parameters of all mixture components in-distinguishable and induces the mode-collapse.



#### Dispersed EM-VAE

$$L(\theta; x) = \text{ELBO} + \beta \cdot L_d,$$

$$L_d = \mathbb{E}_{q_{\phi}(c|x)} A(\boldsymbol{\eta}_c) - \widehat{A}(\mathbb{E}_{q_{\phi}(c|x)} \boldsymbol{\eta}_c).$$

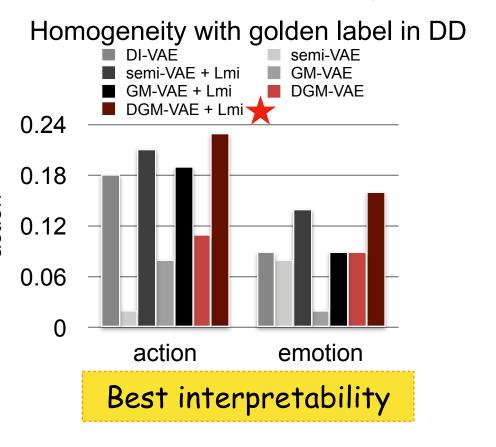
Include an extra positive dispersion term to balance the mode collapse from ELBO

DEM-VAE [W. Shi, H. Zhou, N. Miao, Lei Li, ICML 2020]

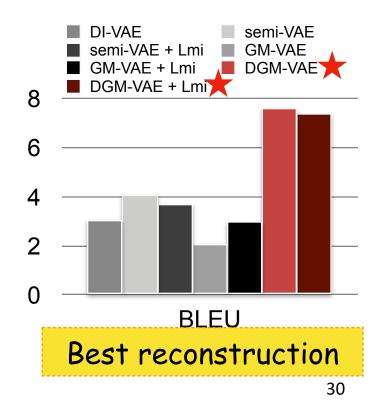
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### **Generation Quality and Interpretability**

## DGM-VAE obtains the best performance in interpretability and reconstruction



BLEU of reconstruction in DD



DEM-VAE [W. Shi, H. Zhou, N. Miao, Lei Li, ICML 2020]

## Generate Sensible Dialog Response with DEM-VAE

#### **Input Context**

Sys: "Taking you to Chevron."

sampling different values of discrete latent variables

(action = thanks)

(action = request-address)

#### **Predict**

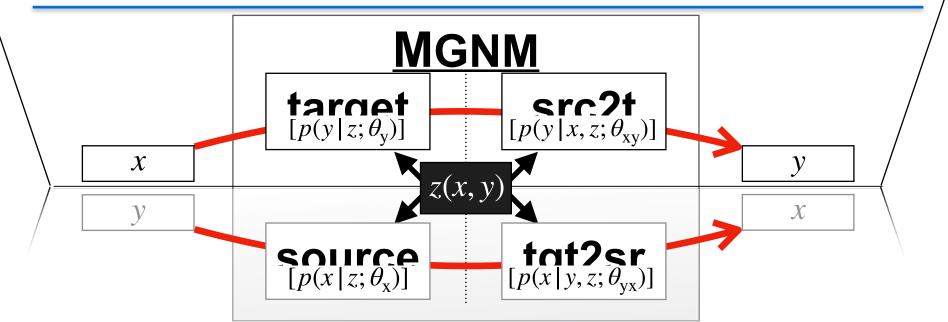
User: "Thank you car, let's go there!"

#### **Predict**

*User:* "What is the address?"

Responses with different actions are generated by sampling different values of discrete latent variables.

## Integrating Four Language Skills with MGNMT

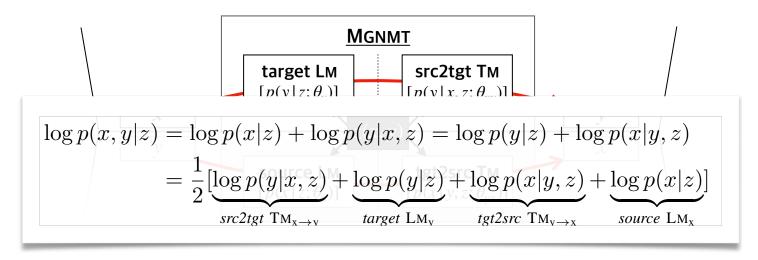


- 1. composing sentence in Source lang
- 2. composing sentence in Target lang
- 3. translating from source to target
- 4. translating from target to source

Benefits
utilizing both
parallel
bilingual data
and nonparallel corpus

### **Approach: Mirror-Generative NMT**

The mirror property to decompose



$$p(x, y | z) = p(y | x, z)p(x | z) = p(x | y, z)p(x | z)$$

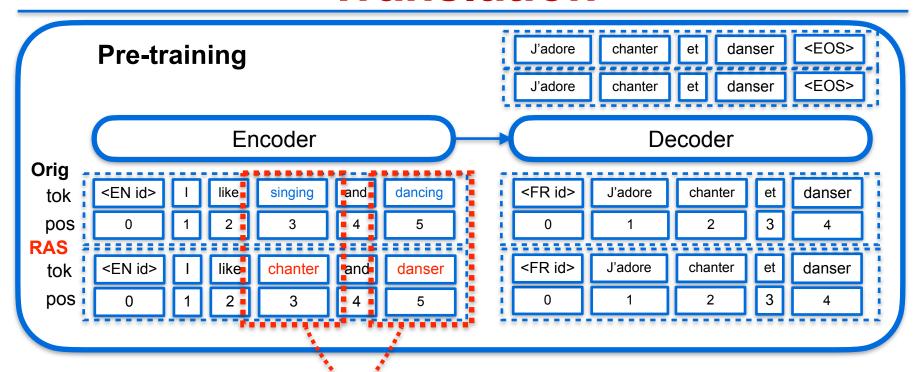
- Relevant TMs & LMs under a unified probabilistic framework!
  - Enables the aforementioned advantages

### MGNMT makes better use of nonparallel data

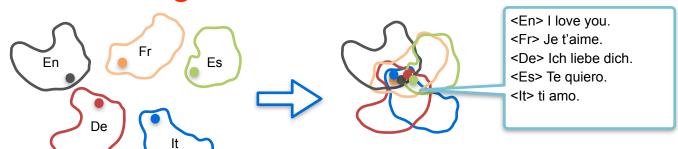
#### Low resource results

	Low-Resource		Cross-Domain			
Model	WmT16 En↔Ro		In-Domain (Ted)		OUT-DOMAIN (NEWS)	
	En-Ro	Ro-En	En-De	DE-EN	En-De	DE-EN
Transformer (Vaswani et al., 2017)	32.1	33.2	27.5	32.8	17.1	19.9
GNMT (Shah & Barber, 2018)	32.4	33.6	28.0	33.2	17.4	20.1
GNMT-M-SSL + non-parallel (Shah & Barber, 2018)	34.1	35.3	28.4	33.7	22.0	24.9
Transformer+BT + non-parallel (Sennrich et al., 2016b)	33.9	35.0	27.8	33.3	20.9	24.3
Transformer+JBT + non-parallel (Zhang et al., 2018)	34.5	35.7	28.4	33.8	21.9	25.1
Transformer+Dual + non-parallel (He et al., 2016a)	34.6	35.7	28.5	34.0	21.8	25.3
MGNMT	32.7	33.9	28.2	33.6	17.6	20.2
MGNMT + non-parallel	34.9	36.1	28.5	34.2	22.8	26.1

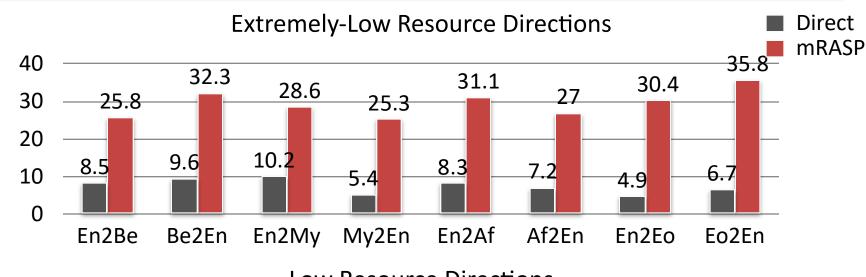
#### mRASP: Multilingual Machine Translation



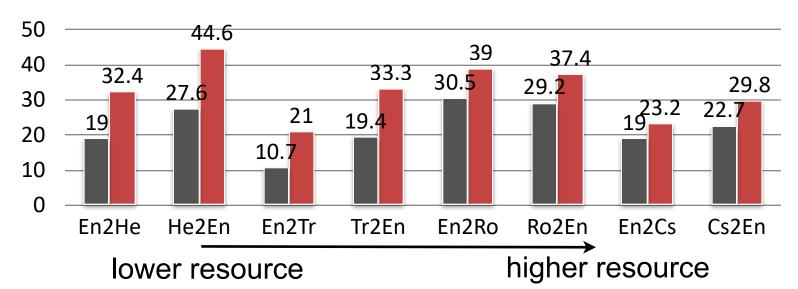
#### Random Aligned Substitution



### mRASP gets universal improvement



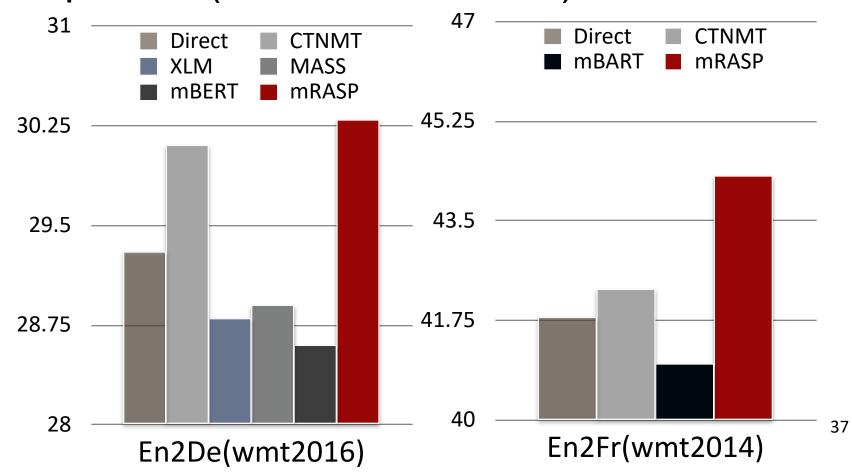
Low Resource Directions



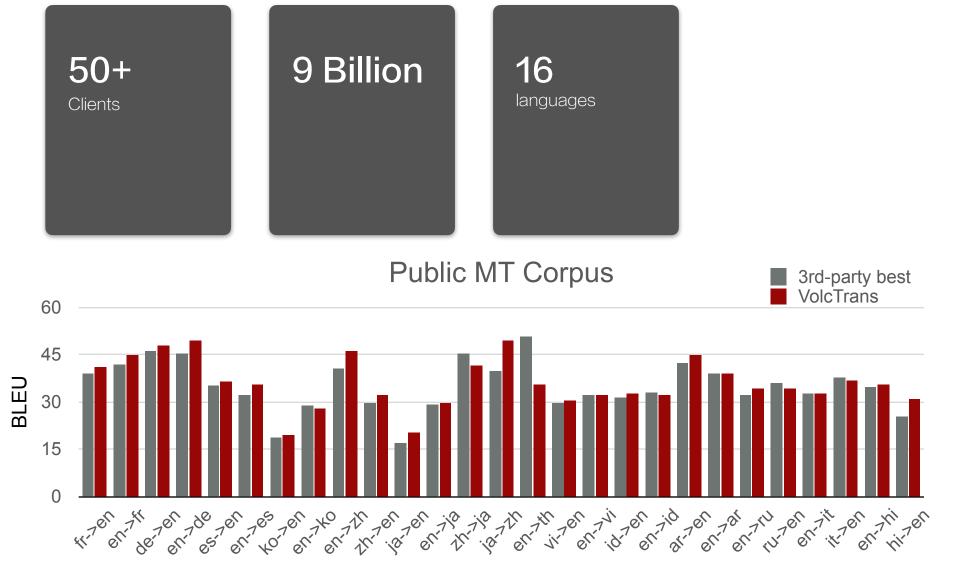
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### mRASP gets universal improvement

 Rich resource benchmarks can be further improved (En->Fr +1.1BLEU).



# VolcTrans fanyi.volcengine.cn



# **Speech-to-Text Translation**



Simultaneous Speech-to-text Translation @ VolcTrans

# Multimodal Machine Writing

Xiaomingbot [R. Xu, J. Cao, M. Wang, J. Chen, H. Zhou, Y. Zeng, Y. Wang, L. Chen, X. Yin, X. Zhang, S. Jiang, Y. Wang, Lei Li, ACL 2020] GraspSnooker [Z. Sun, J. Chen, H. Zhou, D. Zhou, Lei Li, M. Jiang, IJCAI19b]

Jersey Number Recognition with Semi-Supervised Spatial Transformer Network [G. Li, S. Xu, X. Liu, **Lei Li**, C. Wang, CVPR-CVS18]

### **Automatic News Writing in Real-world**

- Tencent: Dreamwriter, started in 2015.9
- Fast Writer Xiaoxin: Xinhuanet, started in 2015.11
- Xiaomingbot: ByteDance, started in 2016.8
- Xiaonan: Southern Weekend, started 2017.1
- Wibbitz: USA Today
- Heliograf: Washington Post

Landon beat Whitman 34-0;

<a href="https://t.co/V6zVPi7a90">https://t.co/V6zVPi7a90</a>
<a href="mailto:@koachkuhn">@koachkuhn</a>
— WashPost HS Sports
<a href="mailto:@WashPostHS">(@WashPostHS)</a>) September 2, 2017



## Xiaomingbot Automatic News Writing System

Winning 2017 Wu Wen-tsün Award in AI from CAAI





北京时间2018年6月23日20时0分,世界杯 G组 第2轮,比利时迎战突尼斯。 最终 比利时5:2战胜突尼斯, 卢卡库,巴舒亚伊,阿扎尔为本队建功 ,哈兹里,布隆为





Post

Thomas Strakosha's 4 saves did not stop Lazio from defeat against Inter Milan, final score 0: 3



Marseille dropped a 0: 2 decision against

PSG in Ligue 1

Sevilla took away a victory against Huesca, 2: 1



**600,000** articles

6 lang

**150,000** followers

### Xiaomingbot: Multilingual Robot News Reporter



#### MULTILINGUAL ROBOT NEWS REPORTER

--- Xiaomingbot ---



# Snooker Commentary Generation Combining Visual Understanding with Strategy Prediction



Balls Detection

Balls' Positions at the Beginning

Red0: (180, 542) Red1: (189, 552) Red2: (179, 555) Red3: (184, 561) Red4: (202, 563)

Red5: (174, 564) Red6: (189, 569) Red11:(197, 590)

Red7: Red12:(241, 595)

Red13:(155, 606)

Red14:(327, 611)

Brown: (183, 163)

Green: (240, 163)

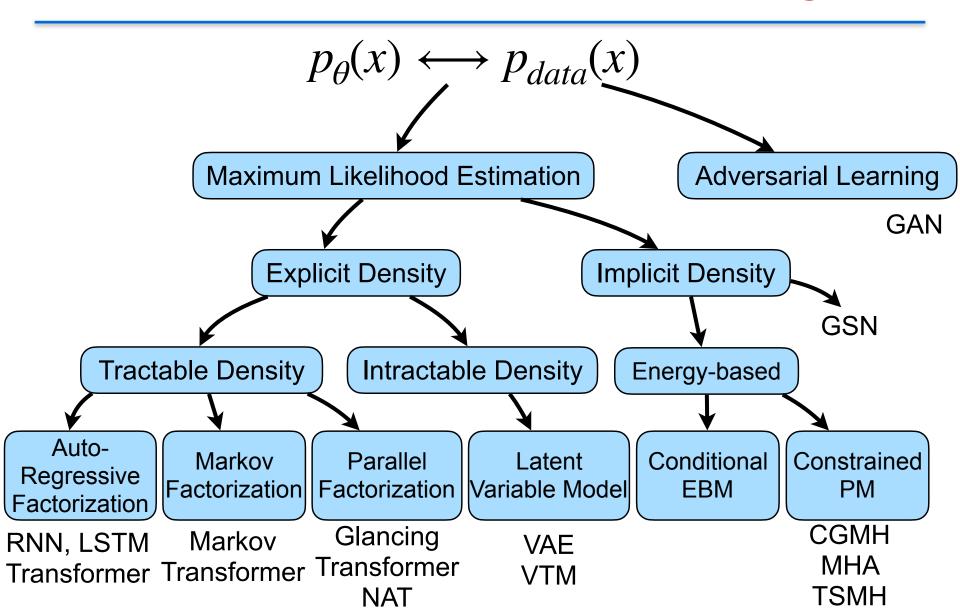
Yellow: (127, 163) Blue: (183, 366

(positions after mapping)

# Summary

- Transformer, LSTM & Softmax: Basic neural generation nets for text
- Disentangled Latent Representation
  - VTM: Learning Latent Templates in Variational Space
  - DSS-VAE: Disentangled syntax and semantic representation
- DEM-VAE: Self identifying meaningful clusters with corpus
- MGNMT:
  - integrate four language capabilities together
  - Utilize both parallel and non-parallel corpus
- Multimodal Machine Writing
  - Xiaomingbot system: 600k articles and 150k followers
- Deployed in multiple online platforms and used by over 100 millions of users

# Recap: DGM Taxonomy



### **Thanks**

- Joint w/ Hao Zhou, Rong Ye, Ning Miao, Wenxian Shi, Zaixiang Zheng, Huangzhao Zhang, Ying Zeng, Jiaze Chen, Han Zhang
- Contact: <u>lileilab@bytedance.com</u>



Multilingual MT Pretraining <a href="https://github.com/linzehui/mRASP">https://github.com/linzehui/mRASP</a>







A high performance sequence processing lib https://github.com/bytedance/lightseq

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