# WakaGPT: Japanese Waka Poem Composer

Ruixuan Tu (ruixuan@cs.wisc.edu)

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## What is Waka?

**Waka** is a traditional Japanese poem with 31 moras in 5-7-5-7-7 pattern.

**Mora:** In linguistics, mora is a basic timing unit in the phonology of some spoken languages; a kana unit in Japanese.

Collection of Hiragana: あいうえおかきくけこさしすせそたちつてとなにぬねのはひふへほまみむめもやゆよらりるれろわをん(one form of kana, cursive)

**Kanji:** Chinese characters used in Japanese writing

Example of Kanji with Hiragana reading: 日本 (にほん, ni ho n) あききぬと/めにはさやかに/見(み)えね とも/風(かぜ)のおとにそ/おとろか れぬる

a ki ki nu to (5) me ni Fa sa ya ka ni (7) mi e ne do mo (5) ka ze no wo to ni zo (7) o do ro ka re nu ru (7)

When autumn came
My eyes clearly
Could not see it, yet
In the sound of the wind
I felt it. Source of Translation

Datasets DO Model

esults

Extensions

# Objective

Given optionally preface sentence, author, and/or leading lines: generate/complete the whole poem.

For example, the text in gray could be generated:

Preface: 秋立つ日よめる

Composed on the first day of autumn.

Author: 敏行 藤原敏行朝臣 (018)

Fujiwara no Toshiyuki (018)

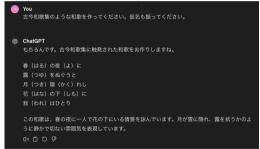
Kana: あききぬと/めにはさやかに/みえねとも/かぜのおとにそ/おとろかれぬる a ki ki nu to/me ni Fa sa ya ka ni/mi e ne do mo/ka ze no wo to ni zo/o do ro ka re nu ru

# Challenges of Waka Generation

#### **Challenges:**

- The poetry is written in classical Japanese, not modern Japanese
- It also requires to generate in the specific format
- No good tokenizer (beyond closed-source 2-gram UniDicS) available
- ChatGPT performs poorly in Waka generation

Ask ChatGPT 4 "Please generate a waka poetry like in Kokinshu collection, and provide the kana reading."



This generated poem is in 5-7-6-6-6, far from the standard 5-7-5-7 pattern.

### **Previous Work**

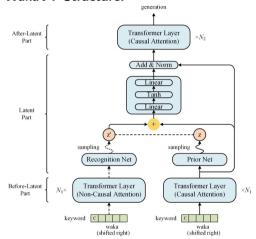
#### WakaVT:

- First waka composer model that use Transformer and VAE
- Trained on same waka dataset

#### Comparison:

- Its task is slightly different from ours: given one keyword, and generate the whole poem
- Our method is simpler without non-causal attention and MLP layers (recognition net and prior net), just the decoder (GPT)
- Our method is based on pre-training on a larger dataset

#### WakaVT Structure:



Novelty of Our Method: Pre-trained Model

### **CHJ** Dataset

- 日本語歴史コーパス CHJ (Classical Japanese Historical Corpus): Nara and Heian Periods (710-1185)
- Row: prior context (19 words), key word, and posterior context (19 words)
- **Size:** 1453767 rows = 3GB CSV files (before tokenization)
- Data Augmentation: different forms of writing (kanji for symbol and kana for pronunciation) to replace the key word
  - Form 0-2 extra copies of each context
  - Learn the reading of non-kana words

**Row Example:** Excerpt from The Kiritsubo Chapter, The Tale of Genji

Prior Context: 更衣 | たち | は | まして | やすから | ず | 。# 朝夕 | の | 宮仕 | に | つけ | て | も | 、 | 人 | の | 心 | を | のみ | Key Word: 動かし (u go ka shi)

Augmentation Candidates: うごかし (u go ka shi), ウゴカシ (u go ka shi)

Posterior Context: |、| 恨み | を | 負ふ | つもり | に | や | あり | けん | 、 | いと | あっしく | なり | ゆき | 、 | もの | 心細げ | に | 里

### Waka Dataset

- 和歌データベース (Waka Poetry Database)
- Row = Prompt Tags (optional Metatags + 1 Poem tag): [詞書
   Meta-Preface] [作者 Meta-Author] [仮名Poem-Kana] [原文 Poem-Original] [整形Poem-Aligned]
- Data Augmentation: different combinations and orders of prompt tags, at most  $(1 \cdot 0! + 2 \cdot 1! + 1 \cdot 2!) \cdot 3 = 15$  lines
- **Size:** 208972 rows = 48.1MB Python Pickle in Pandas before augmentation; 722765 (3.46x) rows after augmentation

#### Preview of df[0]:

>>> d(0)
(\*collection\_date': (延喜五年四月十八日) ((965年5月24日)) , 'collection\_na
me': 古今集', 'collection\_name, 'collection\_name': : 古今集', 'collection\_note': '現存
体に 紅葉書七年や延壽十三年三月十三日の歌も収めらわている。 官職表記は基本十三十
四年成いは延壽十年から十七年までわたるとされたと太連富五年に成立したとしてもその
他とばらくは助射江東が打ちれたとみられる。 'collection\_place': ', 'poem\_autho
r': '元力 在原元方 (169)', 'poem\_comment': '果両資料均番号 :00001', 'poem\_autho
r': '元力 在原元方 (169)', 'poem\_comment': '果両資料均番号 :00001', 'poem\_autho
r': '小力 在原元方 (169)', 'poem\_comment': '果両資料均番号 :00001', 'poem\_autho
r': 'poem\_treace': (円濃) ふるとしじ着をおりも1よめる', 'poem\_text\_type': 'kan
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#### **Augmented Prompt:**

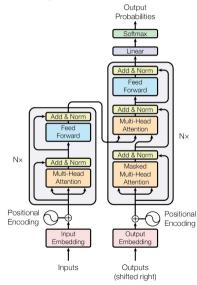
[詞書] ふるとしに春たちける日よめる  $\ n$  [作者] 元方 在原元方  $(169)\ n$  [仮名] としのうちに - はるはきにけり - ひととせを - こそとやいはむ - ことしとやいはむ

# Order of Training

- 1. Kyoto University GPT2: Medium size, character-level
- 2. CHJ-GPT2: one epoch on CHJ Dataset, based on Kyoto University GPT2
- 3. WakaGPT: one epoch on Waka Dataset, based on CHJ-GPT2

Character-level model: no need for tokenization; leverage morpheme understanding on model

Illustration of Transformer Structure: Encoder at left, Decoder at right GPT: Decoder-only Transformer Source: Attention Is All You Need



# Causal LM Training Paradigm

## Pre-training: Self-supervision Objective

Causal LM: predict one token, given all previous tokens but none of the future tokens Self-supervision: treat the input text itself as label, for unlabeled data Generation/Completion: greedy/beam search maximal likelihood tokens one by one Therefore, we predict the token within the input text, and mask the following tokens Equivalent batch size is achieved by dynamic gradient accumulation and gradient

checkpointing on GPUs with lower memory

## Fine-tuning: Prompt Based

**Prompt design:** leave the poem tag at the end of the prompt, so it would be completed based on prior conditions

kono	eiga	ga	kirai	[sos] I	hate	this	movie	[eos

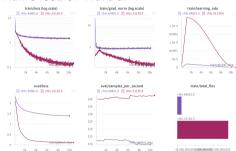
Source: Junjie Hu's CS769

# Hyperparameters

	CHJ-GPT2	WakaGPT		
Train/Test Split	0.9/0.1	0.9/0.1		
Batch Size	32	32		
Number of Train Epochs	1	1		
Warmup Ratio	0.1	0.1		
Learning Rate	3e-4	2e-5		
LR Scheduler	Cosine	Linear		
Weight Decay	0.01	0.01		
Optimizer	AdamW	AdamW		
Adam $\beta_1$	0.9	0.9		
Adam $eta_2$	0.95	0.999		
Adam $\epsilon$	1e-8	1e-8		
DDP Runtime	1d 4h 2m	7h 3m		
GPU	8x A30 24GB (CHTC)	2x A30 24GB (CHTC)		

# Log Highlights

### Run 51130.0: CHJ-GPT2 Run 64801.0: WakaGPT



#### Conclusions

 Training losses of both models converge under regularization

- Not overfitting: train loss and eval loss are very close
- CHJ dataset costs slightly less time to forward than Waka dataset
- However, CHJ dataset is much larger, taken 10x FLOS to train one epoch

#### Limitations

- Underfitting: WakaGPT has non-sufficient (low) learning rate, resulting in larger grad norm and loss than CHJ-GPT2
- Not filtering out augmented pairs for train/eval split

#### Demonstration

The model is hosted as inference endpoints on HuggingFace Spaces and Funix Cloud. You can try it out at either of the following links:

- https://huggingface.co/spaces/TURX/j apanese-Im
- https://gpt.turx.tokyo



## Sample Generation

Prompt: first line あききぬと -

Pronunciation: a ki ki nu to / o mo i ke ru ka na / ya ma za ku ra / sa ki te chi ri nu ru / ha na no shi ta ka ze

Human Assigned Original Form: 秋きぬと思ひけるかな山桜咲きて散りぬる花の下風

Human Translation: When autumn came / I recalled / the mountain cherry blossoms / bloomed and then scattered / flowers and the breeze below.

### Concurrent Work

- CHJ-DeBERTa: Pre-trained masked LM (whole word masking, character-level, and bidirectional attention) on CHJ dataset for non-causal attention/classification tasks
- PageRCNN: object detection for page boundaries from booklet scans based on Faster R-CNN

Demo will be published along with application of these models in this summer.



### **Future Work**

- Fix the limitations mentioned before
- Train on continuum of datasets to adapt variation: from modern, to pre-modern, and finally to classical Japanese
- Seq2Seq applications: recognition, translation, and inference

# Recording

Thanks. Please check for Zoom recording.