Text-to-Lyrics Generation with Image-based Semantics and Reduced Risk of Plagiarism [ISMIR2023]

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Overview

Goal: Text-to-lyrics generation system by suggesting the generated lyrics to users who struggle to find the right words to convey their message.

Obsirable System

- Generate lyrics that convey a message similar to the input text.
- \succ Generate lyrics as a source of new inspiration.

Input Text Driving a car on the seaside Text-to-Lyrics Generation Generated Lyrics

Our Set and S

- > Generate lyrics that are similar in wording to the input text.
- > Generate lyrics that may not provide enough inspiration.

Input Text Driving a car on the seaside Text-to-Lyrics Generation

Generated Lyrics

Anti-Plagiarism for Lyrics Generation

Question: How to check if lyrics are plagiarized?

② Phrases that are safe to generate:

- > New phrases that do not exist in the dataset.
- > Commonly used phrases, such as "I love you".
- Our Phrases with a risk of plagiarizing existing lyrics:
 > Uncommon phrases used in only a few songs.

Proposed Method:

(1) Collect all word N-grams with DF (Document Frequencies) 1-3 in the training data and add them to a list of uncommon phrases, UncommonPhrase.

(2) During the lyrics generation process, if any phrases within the generated lyrics are found in Uncommon-*Phrase*, those lyrics are discarded.

I'm driving in my car Just like a fish on an ocean floor

Driving a car along the coastline

Technical Problem:

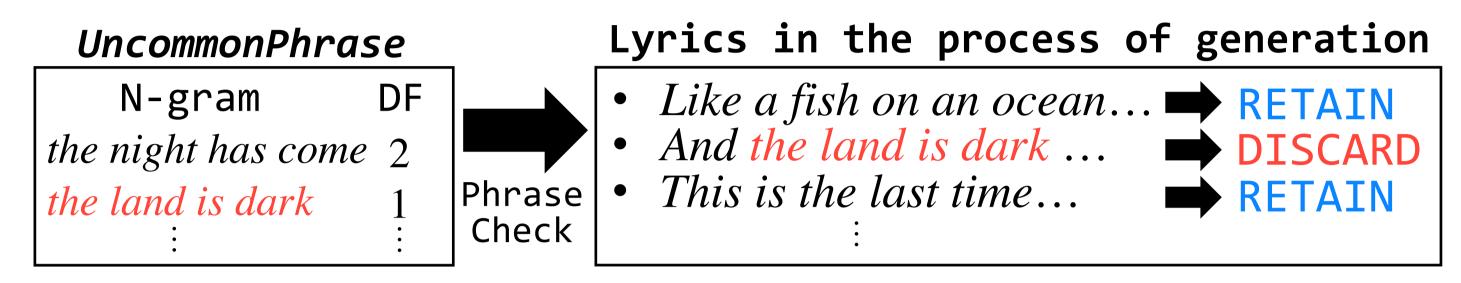
- (1) Training a text-to-lyrics encoder-decoder is not feasible since there is **no text-lyric paired data**.
- (2) Machine learning methods may unintentionally reproduce training data (plagiarize existing lyrics). Experiments

Contributions:

- (1) Propose a two-step pipeline that enables training of a text-to-lyrics generation system even without paired text-lyrics data.
- (2) Propose a lyric generation method that reduces the risk of plagiarizing lyrics from the training data.

Text-to-Lyrics Generation

Key Idea: Generating lyrics from input text by routing through an intermediate image.



To test our lyrics generation, we used **plot summaries** of 20 Disney films from Wikipedia and their theme song lyrics, assuming the lyrics reflect the film's content.

Examples of Generated Lyrics

_	Input Text (plot summary) -	→	Image	\rightarrow	Generated Lyrics
	In a park in England, a young girl named Alice with her cat, Dinah, listens distra- ctedly to her sister's history lesson, and begins daydreaming of a nonsensical wor- ld. She spots a passing White Rabbit				Come to me, come to r I want a little rabbit at it's the same as you

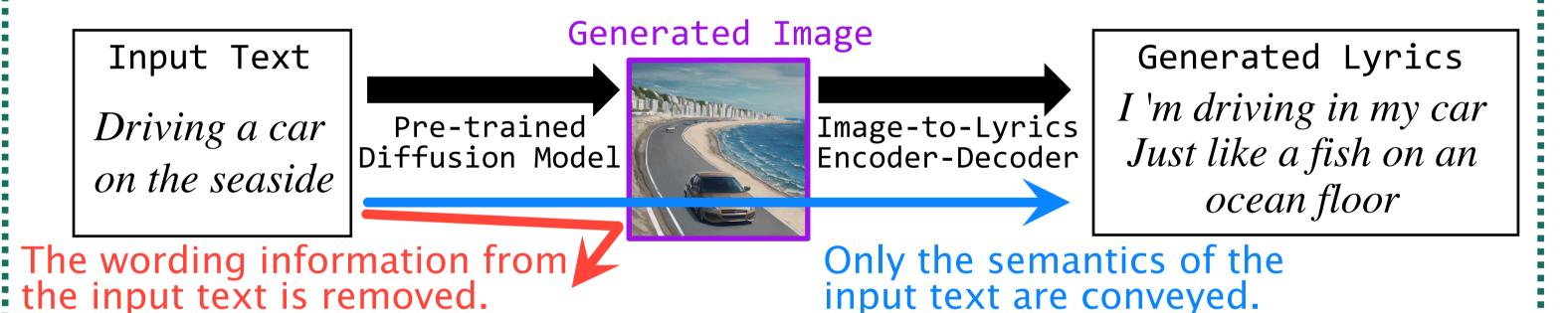
Frozen. Princess Elsa of Arendelle possesses magical powers allowing her to control ice and snow, often using them to play with her younger sister Anna. After Elsa accidentally injures Anna ...

ne to me bbit and ls you



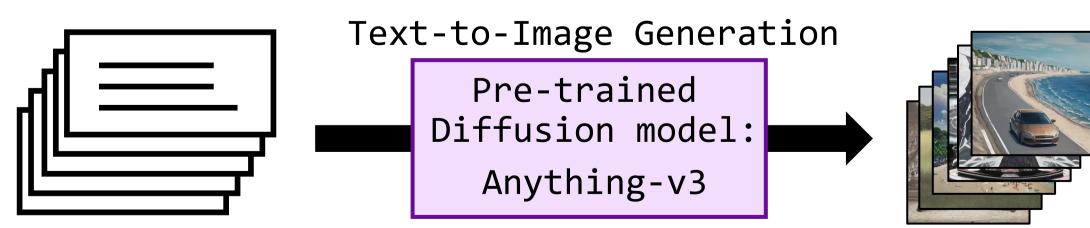
11

12



> Advantages of using the intermediate image:

- The image represents the semantics of the input text without holding its wording information.
- The style of the input text can be any word, phrase, sentence, or paragraph.
- \succ But, we must train the Image-to-Lyrics Enc-Dec: ⊗ To train our Enc-Dec, image-lyric data is required. • We therefore generate images from existing lyrics by using a diffusion model to create training data.



• We succeeded in using images to capture moods of the input text and generating lyrics accordingly.

Quantitative Evaluation:

Perplexity (PPL): Predictability of the phrasing in the original lyrics in the test set.

Normalized Edit Distance (NED): Evaluate whether the proposed method generates lyrics that differ in wording from the input text.

Training paired data for Enc-Dec:	English		Japanese	
[X]-to-Lyrics Generation	PPL↓	NED↑	PPL↓	NED ↑
Image-to-Lyrics (proposed)	84.86	0.78	231.49	0.92
Summary-text-to-Lyrics	346.73	0.69	306.19	0.86
Back-translated-lyrics-to-Lyrics	544.21	0.71	1051.58	0.66
First-half-lyrics-to-Second-half-lyrics	163.98	0.68	583.13	0.90

© Image-lyric pairs are more effective than other paired data sets as training data for Enc-Dec generating lyrics that are semantically related to the input text but differ from it in wording.

Ablation Test for the Anti-Plagiarism Method

927.5K English lyric paragraphs / 1,08M Japanese lyric paragraphs

927.5K generated images/ 1,08M generated images

like

Just

Model Structure: A model that connects a pre-trained Vision Transformer with a Transformer Decoder.

