DALL-E: Text to Image Generation

paper (https://arxiv.org/pdf/2102.12092.pdf)

OpenAI DALL-E 2 announcement (https://openai.com/dall-e-2/)

DALL-E is a generative model for creating Images, conditioned on Text input.

That is:

- Given some Text that describes the output
- DALL-E will create an image based on the Text input

DALL-E: Text to Image

Text input: "An illustration of a baby daikon radish in a tutu walking a dog"

Image output:



Before we began our journey into Generative ML, imagining how to achieve such a result would have been puzzling.

But, in concept, it follows a familiar path

- A Generative Language Model (e.g., GPT) is able to solve a "predict the next token" task
 - Giving a sequence of tokens
 - Generate a token likely to follow
- "Text to Text" as a universal API
 - turn your task into a type of translation of
 - o an input sequence of Source tokens
 - to to an output sequence of tokens

The key prerequisite to DALL-E (in addition to a Language Model) • representation of a (flattened) Image as a sequence of (discrete) tokens

We have already encountered such a representation

- A Discrete Variational Autoencoder (dVAE) such as the VQ-VAE represents an Image
- as a higher dimensional Tensor (e.g, 2D grid)
- of discrete values
 - indexes into a finite "codebook" of latent vectors
 - the codebook implements an embedding of Image "elements" into a finite set of vector encodings

Hopefully: the idea underlying DALL-E has come into focus

- We train a Language Model on a sequences of generalized tokens
 - First part of sequence are Text tokens
 - Separated by an "end of Text" token
 - Second part of sequence are Image tokens
- The Language Model learns
 - not only what "words" follow the prefix of a sequence of word tokens
 - but what Image tokens follow the "end of Text"

Voila!

Text to Image is just another use of the "Text to Text" API

• akin to translating from one language to another

Details

A dVAE is trained

- ullet Encoder reduces a 3D (RGB) image to smaller (32 imes 32) spatial grid of latent vectors
- the dVAE learns a codebook
 - of 8192 elements (the Image Vocabulary size)
- Using the dVAE Encoder, an image is translated into a (32 imes 32) spatial grid

A Text Encoder is used

- BPE encoding of tokens
- $\bullet \ \ {\sf Text \, Vocabulary \, size \, is \, 16,384}$

The Language Model is trained

- Examples are (Text, Image) pairs
 - Obtained from images with captions
- Given training example (text⁽ⁱ⁾, image⁽ⁱ⁾)
- ullet Text $\operatorname{text}^{(i)}$ is converted to a sequence using the Text Encoder
 - maximum text sequence length limited to 256 Text tokens
- Image $\mathrm{image^{(i)}}$ is converted to a (32×32) spatial grid
 - lacktriangle which is flattend to a sequence of 1024=32*32 Image tokens

The training example is thus a sequence of

- 256 Text tokens
- $\bullet~1024~{\rm Image~tokens}$

The Language Model learns to Autoregressively model the training sequences

• Using Decoder style Transformer

Here is what a training example that mixes text and an image looks like:

Training example: Representing mixed Text + Image as a sequence



Figure 10. Illustration of the embedding scheme for a hypothetical version of our transformer with a maximum text length of 6 tokens. Each box denotes a vector of size $d_{\rm model}=3968$. In this illustration, the caption has a length of 4 tokens, so 2 padding tokens are used (as described in Section 2.2). Each image vocabulary embedding is summed with a row and column embedding.

Notice the positional embeddings

- position in text
- row/column position for image

A Language Model is not deterministic: running the same input multiple times can generate different outputs

- LLM creates a probability distribution over token vocabulary
 - we sample a token from this distribution as the LLM "next token" output

DALL-E uses CLIP to rank the outputs

• CLIP finds the images that are the least distance from the Text

Interesting results and observations

OpenAI announcement of DALL-E 2 (https://openai.com/dall-e-2/)

DALL-E seems to have learned to generalize

- the "baby daikon radish wearing a tutu walking a dog"
 - even though radishes don't wear clothing, DALL-E positions the tutu around the "waist" of the radish
 - the radish is in a ballerina (dancer who wears a tutu) pose
 - the dog is on a leash (as are dogs being walked)
 - even though a radish does not have arms, DALL-E creates an arm to hold the leash
- Different styles of Image are created according to the text "illustration", "sketch", "photo", "painting in the style of Picasso"

Image editing via Text

- Given the text sequence "the exact same cat on the top as a sketch at the bottom"
- Followed by Image tokens (half the length of the total Image sequence length) encoding a cat
- Will generate the remaining available Image tokens of the cat in the style of a sketch

Doesn't always understand

• "tree bark" (the "skin" of a tree) is **not** an animal barking at a tree

Social concerns

The DALL-E 2 preview included a <u>model card (https://github.com/openai/dalle-2-preview/blob/main/system-card 04062022.md#bias-and-representation)</u> describing possible Risks and Limitations.

Among them are

- <u>Inappropriate training examples (https://github.com/openai/dalle-2-preview/blob/main/system-card 04062022.md#model-training-data)</u>
- "Signing" to indicate DALL-E generated image (https://github.com/openai/dalle-2-preview/blob/main/system-card_04062022.md#signature-and-image-provenance)
- <u>Biases present in training data (https://github.com/openai/dalle-2-preview/blob/main/system-card_04062022.md#bias-and-representation)</u>

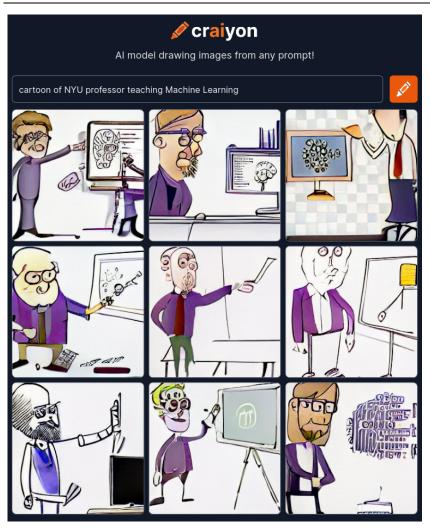
Playing with DALL-E

There is a <u>waiting list (https://labs.openai.com/waitlist)</u> to get access to the official version of DALL-E.

However there are some Open Source replicas (trained on much less data)

- Craiyon (https://craiyon.com)
 - Advertisement supported :-(
- Huggingface.co (https://huggingface.co/spaces/dalle-mini/dalle-mini)
 - Moving to Craiyon

Image output:



- It gets the NYU colors right!
 - Princeton professors wear orange

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In [1]: print("Done")
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Done