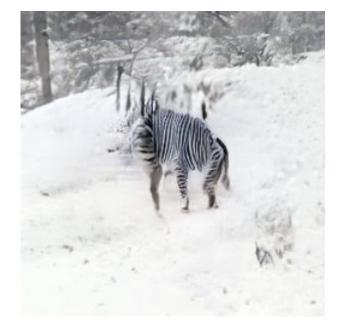
5/5 進度報告

應名宥

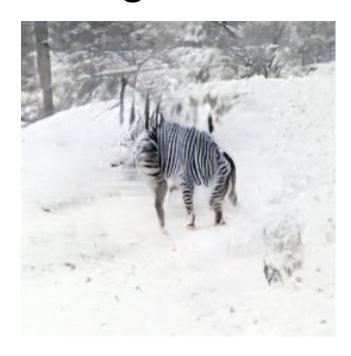
text - image similarity





score : 35.16 score : 40.59

add image realistic score



only clip rating

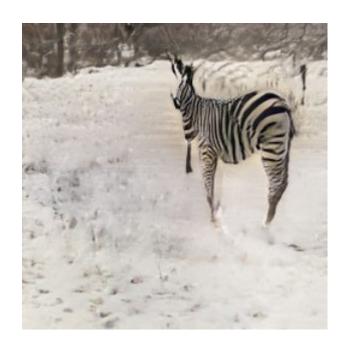


clip rating and vgg19 score

NSGA2



rand 1000 sample



population: 100 #offspring: 25 generation: 50

win





a cat is lay on the table.

result of failure





a street scene with a double-decker bus on the road.

166.59

151.25

analysis





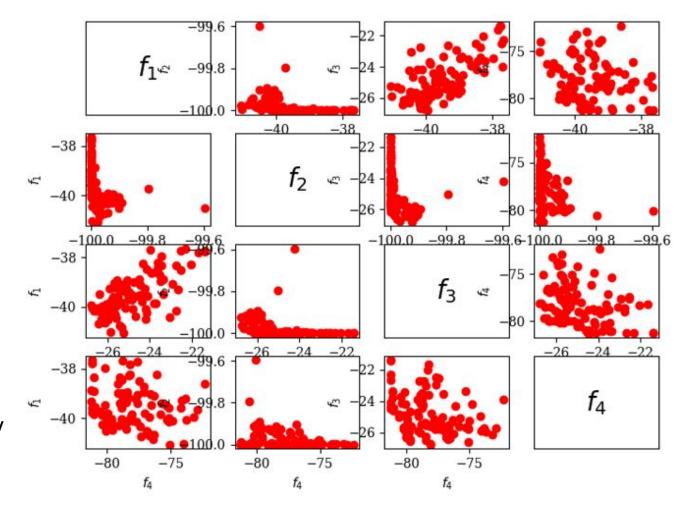
```
txt = a street scene with a double-decker bus on the road.
description - text similarity score = 35.97
image realistic score = 89.17
predict label similarity score = 26.11
label = trolleybus , probability = 0.33
label = streetcar , probability = 0.25
label = minibus , probability = 0.17
label = passenger car , probability = 0.11
label = amphibian , probability = 0.04
```

```
txt = a street scene with a double-decker bus on the road.
description - text similarity score = 38.34
image realistic score = 99.82
predict label similarity score = 28.43
label = trolleybus , probability = 0.84
label = streetcar , probability = 0.14
label = minibus , probability = 0.01
label = passenger car , probability = 0.01
label = school bus , probability = 0.00
```

a zebra is walking in the snow.

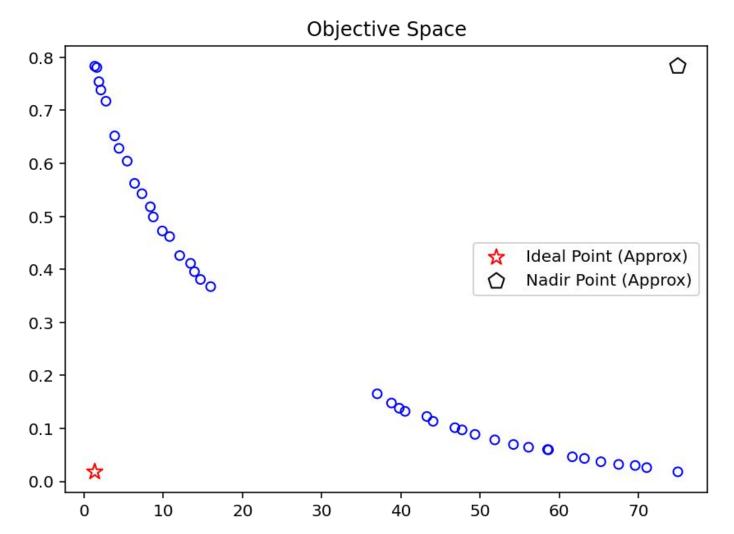
population: 100 #offspring: 100 generation: 50

- 1. text-image-simularity
- 2. image realistic score
- 3. image-label-simularity
- 4. word-label-simularity



MCDM

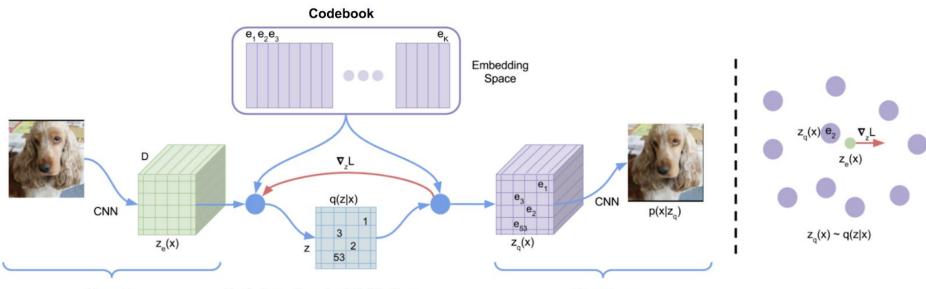
```
def mcdm(res = None):
    F = res.F
    approx ideal = F.min(axis=0)
    approx nadir = F.max(axis=0)
   # normalization
    nF = (F - approx ideal) / (approx nadir - approx ideal)
    # setting weigth
    weights = np.array([0.30, 0.40, 0.30])
    decomp = get decomposition("asf")
    # inverse
    I = decomp.do(nF, 1/weights).argmin()
    print("Best regarding decomposition: ASF {} - {}".format(I, -F[I]))
    return I
```



explain

- Normalizing the obtained objective values regarding the boundary points.
- you might be wondering why the weights are not passed directly, but 1/weights. For ASF, different formulations exist, one where the values are divided and one where they are multiplied. In pymoo, we divide, which does not reflect the idea of the user's criteria. Thus, the inverse needs to be applied.

noise codebook



Encoder

Posterior categorical distribution:

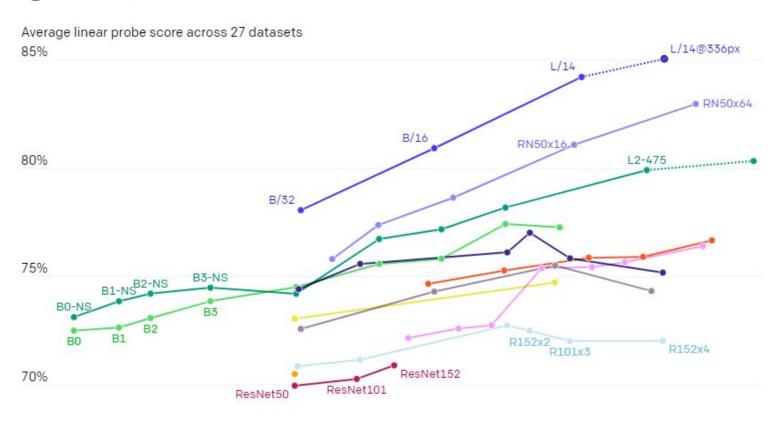
Decoder

$$q(\mathbf{z} = \mathbf{e}_k | \mathbf{x}) = \begin{cases} 1 & \text{if } k = \arg\min_i \|\mathbf{z}_e(\mathbf{x}) - \mathbf{e}_i\|_2 \\ 0 & \text{otherwise.} \end{cases}$$

codebook benefit

- 範圍可控制
- 穩定
- 生成結果更好

large clip model



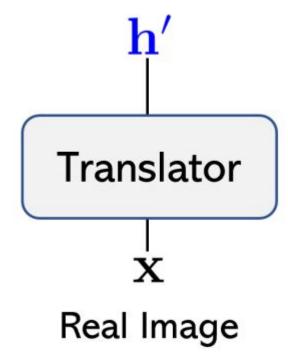
datasets

- COCO2017
- Image-net (classification)

```
templates = [
    'a photo of a {}.',
    'a photo of the {}.',
]
```



pseudo text feature



Diffusion Models Beat GANs on Image Synthesis

Prafulla Dhariwal*
OpenAI
prafulla@openai.com

Alex Nichol* OpenAI alex@openai.com

ref: https://arxiv.org/pdf/2105.05233.pdf

study

- CLIP
- VQGAN (Taming Transformers)
- Diffusion Models Beat GANs on Image Synthesis
- Guided Diffusion
- Make-A-Scene