

VAE Variational Autoencoder バイアスエンコーダー

- VAEの構成要素はEncoderとDecoder

Encoder(AE)とVAE

Encoder(AE)

- 生成モデル
- 判別モデル
- 入力 $x \rightarrow \text{Encoder} \rightarrow \text{Latent}$
- 出力

VAE

- VAEの構成要素
- 生成過程($P(x|z)$) $\mu(z), \sigma(z)$
- 判別過程($P(z|x)$)
- VAEの特徴

Latent

- $P(x)$
- $P(z|x)$
 - $P(z|x)$
- VAEの構成要素 **P(x)**
 - $P(x) = \int P(z)P(x|z)$
 - $P(z|x) \sim N(0, I)$
 - $P(z|x)$
 - $P(x|z)$



- Decoder** $P(x|z)$ (NN) $P(x|z) \sim \mu(z) + \sigma(z)$
 $\mu(z) \times \sigma(z)^2 P(x|z)$
- Encoder** $P(z|x)$ (NN') $P(z|x) \sim q(z|x) \sim \mu'(x) + \sigma'(x)$
 $\mu'(x) \times \sigma'(x)^2 P(z|x)$



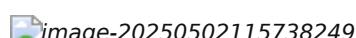
Latent

Latent variable [1. 潜在変数Variational Autoencoder — 潜在変数の定義](#)

X観測変数 **Z**潜伏変数 **Unobserved variable**

X観測変数 **Z**潜伏変数

- $P(X|Z)$
- $P(Z|X)$



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$$\mathcal{D} = \{(z^{\{1\}}, x^{\{1\}}), \dots, (z^{\{N\}}, x^{\{N\}})\}$$


$$\mathcal{L}(\theta; \mathcal{D}) = \sum_{i=1}^N p_\theta(x^{\{i\}} | z^{\{i\}})$$


$$\ln p_\theta(x^{\{i\}} | z^{\{i\}}) = \sum_{j=1}^N \ln \int_z p_\theta(x^{\{i\}}, z) dz$$


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\$jenson\$ \$ \$ \text{f(x)} \$ | f(\mathbb{E}[X]) \geq \mathbb{E}[f(X)] | \text{f(x)} \$ \$ \$ \mathbf{Z} \$ q_{\phi}(z|x)\$
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\begin{aligned}
& \mathcal{L}(\theta; x) = \ln p_{\theta}(x) + \int_z q_{\phi}(z|x) \ln \frac{p_{\theta}(x,z)}{q_{\phi}(z|x)} dz \\
&= \ln \mathbb{E}_{\phi}[q_{\phi}(z|x)] - \mathbb{E}_{\phi}[\ln \frac{p_{\theta}(x,z)}{q_{\phi}(z|x)}]
\end{aligned}

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\$\$ \begin{aligned} \mathcal{L}(q,\theta) &= \mathbb{E}[z \sim q(\phi)] [\ln p_\theta(x,z)] - \\ &\quad \mathbb{E}[z \sim q(\phi)] [\ln q_\phi(z)] \quad \&= \mathbb{E}[z \sim q(\phi)] \\ &[\ln p(z) + \ln p_\theta(x|z)] - \mathbb{E}[z \sim q(\phi)] [\ln q_\phi(z)] \quad \&= \mathbb{E}[z \sim q(\phi)] [\ln p(z)] + \mathbb{E}[z \sim q(\phi)] [\ln p_\theta(x|z)] - \mathbb{E}[z \sim q(\phi)] [\ln q_\phi(z)] \quad \&= \mathbb{E}[z \sim q(\phi)] [\{\ln p_\theta(x|z)\}] + \underbrace{\mathbb{E}[z \sim q(\phi)] [\ln p(z)]}_{[\ln q_\phi(z)]} \{\text{KL}(q_\phi(z)||p(z))\} \quad \&= \mathbb{E}[z \sim q(\phi)] [\ln p_\theta(x|z)] - \underbrace{\text{KL}(q_\phi(z||p(z))}_{q_\phi(z)\text{text}\{\text{KL}(q_\phi(z||p(z))\}} p(z)\text{text}\{\text{KL}(q_\phi(z||p(z))\}} \end{aligned} \$$

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$$\mathcal{L}(q, \theta) = \ell(\theta; x) = \underbrace{\mathbb{E}_{\phi}[\ell(\phi(x); x)]}_{= q_{\phi}(x)} + \mathbb{E}_{\phi}[\ell(\phi(x); x)]$$

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q{\phi}(z|x)\} \{ \{ \ln p_{\theta}(x|z) \} \} \{ \text{reconstruction term} \} \} \underbrace{KL(q(\phi|z|x) | p(z))}_{\text{prior matching term}} \&= \ell(\theta; x) \end{align} $$

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□□	□□	□□□□	□□□□
\$p(z)\$	□□□□□□□□	□□□□□□\$N(0,I)	□
\$q_{\{\backslash phi\}}(z	\$x)	□□□□□	□□□□□\$ \backslash phi\$
\$p_{\{\backslash theta\}}(x	\$z)	□□□□□	□□□□□\$ \backslash theta\$
\$p_{\{\backslash theta\}}(z	\$x)	□□□□□	□□□□□□□□□□□
\$p_{\{\backslash theta\}}(x)\$	□□□□□□	□□□□□□□□□□□	□□VAE

- $p(z)$ \rightarrow $p(z|y)$
 - $p(z|x)$ \rightarrow $p(z|y)$
 - $p(z|y) \approx p(z|x)$ \rightarrow $p(z|y) \approx p(z|x)$
 - $p(z|y) \approx p(z|x)$ \rightarrow **ELBO** \rightarrow $p(z|y) \approx p(z|x)$

100

$q_{\{\phi\}}(z|x) = \mathcal{N}(\mu_z, \Sigma_z)$ μ_z Σ_z

KL-□-□

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## \begin{align} KL(q_{\phi}(z) || p(z)) &= KL(\mathcal{N}(\mu_z, \Sigma_z) || \mathcal{N}(0, I)) \&= \\
&\frac{1}{2} (\text{tr}(\Sigma_z) + \mu_z^\top \mu_z - k - \log \det(\Sigma_z)) \end{align} ##
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$\vdash \exists x \exists y \exists z \exists w \exists v \exists u \exists t \exists s \exists r \exists p \exists q \ \{\phi(z|x)\} \rightarrow \psi(w|y)$

100

$\underbrace{\mathbb{E}\{z \mid \sim q(\phi)(z|x)\}[\ln p_\theta(x|z)]}_{\text{(reconstruction term)}}$
 $\approx \mathbb{E}_{\mu_x}[z \mid \sim q(\phi)(z|x)]$
 $\mathbb{E}_{\mu_x}[z \mid \sim q(\phi)(z|x)] = \mu_x$
 $\mu_x = \text{decoder}_\theta(z)$
 $q(\phi)(z|x) \approx \frac{1}{L} \sum_{l=1}^L \ln p_\theta(x|z^l)$
 $L = 1$
 VAE *reparameterization trick*

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A horizontal row of 20 empty square boxes for writing responses.

$\mathcal{N}(\mu_z, \Sigma_z) \mathcal{N}(0, I)$

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\mu\$ \epsilon\$ \sim \mathcal{N}(0, I)\$ \mu\$ \epsilon\$ \sim \mathcal{N}(\mu_z, \Sigma_z)\$ \begin{aligned} z &= \mu_z + \sqrt{\Sigma_z} \cdot \epsilon \\ &= \mu_{\theta}(x) + \sqrt{\Sigma_{\phi}(x)} \cdot \epsilon \end{aligned}

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## \mathcal{L}(q,\theta) = \underbrace{\{\mathbb{E}[z \sim q(\phi)(z|x)]\{\ln p_\theta(x|z)\}}_{\text{(reconstruction term)}} - \underbrace{KL(q(\phi)(z|x)||p(z))}_{\text{(prior matching term)}} ##
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1. $\mu\sigma$
 2. $\mu\sigma$
 3. $\mu\sigma$
 4. $\mu\sigma$
 5. $\mu\sigma$
 6. $\mu\sigma$
 7. $\mu\sigma$

AE Auto-encoder

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TensorFlow.js چیست و چه کاربردهایی دارد؟

TensorFlow.js

TensorFlow.js چیست و چه کاربردهایی دارد؟

TensorFlow.js

TensorFlow.js چیست و چه کاربردهایی دارد؟

1. چیزی که Encoder است

- "TensorFlow.js" را "Tensor" می‌نامند
- "Tensor" است
- "Tensor" است

2. چیزی که Latent Space است

- "Tensor" است
- "Tensor" است
- "Tensor" است

3. چیزی که Decoder است

- "Tensor" است
- "Tensor" است
- "Tensor" است

Tensor

1. چیزی که Encoder است

2. چیزی که Encoder است 784 × 32
3. چیزی که Encoder است
4. چیزی که Encoder است "Tensor" است
5. چیزی که Encoder است

Tensor

• "Tensor"

• "Tensor"

• "Tensor"

• "Tensor"

• "Tensor"