

2.3 連続確率分布.

* \wedge -分布.

$\mu \in (0, 1)$: 変数 μ 生成.

pdf. $\text{Beta}(\mu | a, b) = C_B(a, b) \mu^{a-1} (1-\mu)^{b-1} \quad (a, b \in \mathbb{R}_{>0})$

$$\int_0^1 \text{Beta}(\mu | a, b) d\mu = 1 \quad \text{より}$$

$$C_B(a, b) \int_0^1 \mu^{a-1} (1-\mu)^{b-1} = 1.$$

\downarrow \wedge -分布の定義.

$$C_B(a, b) B(a, b) = 1. \quad \therefore C_B(a, b) = \frac{1}{B(a, b)}$$

$$B(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)} \quad \text{という性質がある.}$$

対数表示:

$$\log \text{Beta}(\mu | a, b) = (a-1) \log \mu + (b-1) \log (1-\mu) + \log C_B(a, b)$$

平均.

$$\begin{aligned} \mathbb{E}[X] &= \int_0^1 x C_B(a, b) x^{a-1} (1-x)^{b-1} dx \\ &= C_B(a, b) \int_0^1 x^a (1-x)^{b-1} dx = C_B(a, b) B(a+1, b) \\ &= \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \frac{\Gamma(a+1)\Gamma(b)}{\Gamma(a+b+1)} = \frac{\Gamma(a+b) \cdot a\Gamma(a)}{\Gamma(a) \cdot (a+b)\Gamma(a+b)} = \frac{a}{a+b}. \end{aligned}$$

\rightarrow ガンマ関数の性質
 $\Gamma(x+1) = x\Gamma(x).$

$$E[\log X] = C_B(a,b) \int_0^1 (\log x) x^{a-1} (1-x)^{b-1} dx,$$

$$E = 32^a,$$

← ベータ関数の形にもっていきたいけど
log x が邪魔.

$$\frac{d}{da} x^a = \frac{d}{da} (e^{\log x})^a = \frac{d}{da} e^{a \log x}$$

$$= e^{a \log x} \cdot \log x$$

$$= x^a \log x.$$

← 微分の定義

$$f'(a) = \lim_{h \rightarrow 0} \frac{f(a+h) - f(a)}{h}$$

← 互換してもOK. わんどういって

これを適用すると,

$$E[\log X] = C_B(a,b) \int_0^1 \frac{\partial}{\partial a} x^{a-1} (1-x)^{b-1} dx$$

$$= C_B(a,b) \frac{\partial}{\partial a} \int_0^1 x^{a-1} (1-x)^{b-1} dx$$

↓ 微分と積分の交換は
ここでOK.

$$= C_B(a,b) \frac{\partial}{\partial a} \left(\frac{1}{C_B(a,b)} \int_0^1 \text{Beta}(x|a,b) dx \right)$$

$$= C_B(a,b) \frac{\partial}{\partial a} \left(\frac{1}{C_B(a,b)} \right) = \frac{1}{B(a,b)} \frac{\partial}{\partial a} B(a,b)$$

$$= \frac{\partial}{\partial a} (\log B(a,b))$$

$$\downarrow \frac{d}{dx} \log f(x) = \frac{\frac{d}{dx} f(x)}{f(x)}$$

$$= \frac{\partial}{\partial a} (\log \Gamma(a) + \log \Gamma(b) - \log \Gamma(a+b))$$

$$= \psi(a) - \psi(a+b),$$

$$\left(\psi(x) = \frac{d}{dx} \log \Gamma(x) \right)$$

同様にして,

$$E[\log(1-X)] = \psi(b) - \psi(a+b).$$

・ $\mathcal{I} = \text{tot}^0 -$

$$\begin{aligned} H[\text{Beta}(\mu|a,b)] &= -\mathbb{E}[\log \text{Beta}(\mu|a,b)] \\ &= -(a-1)\mathbb{E}[\log \mu] - (b-1)\mathbb{E}[\log(1-\mu)] - \log C_B(a,b) \\ &= -(a-1)\psi(a) - (b-1)\psi(b) + (a+b-2)\psi(a+b) - \log C_B(a,b). \end{aligned}$$

* Dirichlet/分布.

$\pi = \begin{pmatrix} \pi_1 \\ \vdots \\ \pi_k \end{pmatrix}$ k -dim. Prob. vec. を生成.

pdf. $\text{Dir}(\pi|\alpha) = C_D(\alpha) \prod_{k=1}^k \pi_k^{\alpha_k-1}$. $\alpha > 0$.

$$C_D(\alpha) = \frac{\Gamma(\sum_{k=1}^k \alpha_k)}{\prod_{k=1}^k \Gamma(\alpha_k)}$$

◦ Dirichlet/分布が確率分布であることの確認.

π : prob. vec. ならば, $\pi_i \geq 0$ ($i=1, \dots, k-1$) が決まると,

$\pi_k = 1 - \sum_{i=1}^{k-1} \pi_i$ が決定される. よって,

$D_k := \{ \pi \in \mathbb{R}^k \mid \pi_i \geq 0 \ (i=1, \dots, k), \sum_{i=1}^k \pi_i \leq 1 \}$ とし

D_{k-1} で $\text{Dir}(\pi|\alpha)$ を積分する. $d\pi_1 \dots d\pi_{k-1} = d\pi_{1:k-1}$ だと $\pi_k < 1$

$$\int_{D_{k-1}} \text{Dir}(\pi|\alpha) d\pi_{1:k-1} = \frac{\Gamma(\sum_{k=1}^k \alpha_k)}{\prod_{k=1}^k \Gamma(\alpha_k)} \int_{D_{k-1}} \prod_{k=1}^k \pi_k^{\alpha_k-1} d\pi_{1:k-1} .$$

あと, 以下の主張を示せばよい.

$$[\text{Claim}] \int_{D_{K-1}} \prod_{k=1}^K \tau_k^{\alpha_k-1} d\tau_{1:k-1} = \frac{\prod_{k=1}^K \Gamma(\alpha_k)}{\Gamma(\sum_{k=1}^K \alpha_k)}.$$

$$\text{pf. } \prod_{k=1}^K \Gamma(\alpha_k) = \prod_{k=1}^K \int_0^\infty t_k^{\alpha_k-1} e^{-t_k} dt_k$$

$$= \int_0^\infty \dots \int_0^\infty \prod_{k=1}^K (t_k^{\alpha_k-1} e^{-t_k}) dt_k.$$

こゝで変数変換 $(t_1, \dots, t_k) \rightarrow (\tau_1, \dots, \tau_{k-1}, \alpha)$ を次のようにする:

$$t_i = \tau_i \alpha \quad (i=1, \dots, k-1), \quad t_k = \left(1 - \sum_{i=1}^{k-1} \tau_i\right) \alpha. \quad (*)$$

$$1 - \sum_{i=1}^{k-1} \tau_i =: \tau_k \text{ とおく.}$$

積分範囲 $[0, \infty)^k$ がどのように変換されるかを見る:

$$t_i \geq 0 \quad (i=1, \dots, k) \text{ より}$$

$$\tau_i \geq 0 \text{ and } 1 - \sum_{i=1}^{k-1} \tau_i \geq 0 \text{ and } \alpha \geq 0 \text{ と考えられる.}$$

$$\left(\begin{array}{l} \tau_i \leq 0 \text{ and } 1 - \sum_{i=1}^{k-1} \tau_i \leq 0 \text{ and } \alpha \leq 0 \text{ も考えられるが,} \\ \text{改め } \tau_i' = -\tau_i, \quad 1 - \sum_{i=1}^{k-1} \tau_i' = -\left(1 - \sum_{i=1}^{k-1} \tau_i\right), \quad \alpha' = -\alpha \\ \text{と変数変換すればこの Jacobian の絶対値は 1 なの OK.} \end{array} \right)$$

ゆゑ、積分範囲は、 $\mathcal{S} := \underbrace{D_{K-1}}_{\tau_i \geq 0} \times \underbrace{[0, \infty)}_{\alpha}$ となる。

また、変数変換(*)の Jacobian は次のとおり。

$$J(\pi_1, \dots, \pi_{k-1}, x) = \det \frac{\partial(t_1, \dots, t_k)}{\partial(\pi_1, \dots, \pi_{k-1}, x)}$$

$$= \det \left(\begin{array}{ccc|c} x & & & \pi_1 \\ & x & & \pi_2 \\ & & \ddots & \vdots \\ 0 & & & \pi_{k-1} \\ \hline -x & -x & \dots & -x \\ & & & \pi_k \end{array} \right) = x^{k-1} (\geq 0).$$

$$\therefore \prod_{k=1}^k \Gamma(\alpha_k)$$

$$= \int_{\mathcal{D}} \prod_{k=1}^k \left((\pi_k x)^{\alpha_k - 1} e^{-\pi_k x} \right) \underbrace{x^{k-1}}_{\text{Jacobian の 絶対値}} d\pi_{1:k-1} dx$$

$$= \int_{\mathcal{D}_{K-1}} \prod_{k=1}^k \pi_k^{\alpha_k - 1} d\pi_{1:k-1} \times \int_0^\infty x^{\sum_{k=1}^k \alpha_k - k + k - 1} e^{-\sum_{k=1}^k \pi_k x} dx$$

$$= \int_{\mathcal{D}_{K-1}} \prod_{k=1}^k \pi_k^{\alpha_k - 1} d\pi_{1:k-1} \times \Gamma\left(\sum_{k=1}^k \alpha_k\right).$$

対数表示:

$$\log \text{Dir}(\pi | \alpha) = \sum_{k=1}^k (\alpha_k - 1) \log \pi_k + \log C_D(\alpha).$$

平均.

$$E[\pi] = \int \pi C_D(\alpha) \prod_{k=1}^k \pi_k^{\alpha_k - 1} d\pi_{1:k-1}$$

第k要素に注目.

$$E[\pi_k] = \int \pi_k C_D(\alpha) \prod_{k'=1}^k \pi_{k'}^{\alpha_{k'} - 1} d\pi_{1:k-1}$$

$$= C_D(\alpha) \int \pi_1^{\alpha_1-1} \dots \pi_{k-1}^{\alpha_{k-1}-1} \pi_k^{\alpha_k} \pi_{k+1}^{\alpha_{k+1}-1} \dots \pi_K^{\alpha_K-1} d\pi_{1:k-1}$$

$$= \frac{C_D(\alpha)}{C_D(\alpha_k)} \int \text{Dir}(\pi | \alpha_k) d\pi_{1:k-1} \quad (\alpha_k = \alpha + e_k)$$

$$= \frac{\prod_{k=1}^K \Gamma(\alpha_k)}{\prod_{k=1}^K \Gamma(\alpha_k + 1)} = \frac{\alpha_k}{\sum_{k=1}^K \alpha_k}$$

$$\therefore \mathbb{E}[\pi_k] = \frac{1}{\sum_{k=1}^K \alpha_k} \alpha_k$$

$\log \pi = \begin{pmatrix} \log \pi_1 \\ \vdots \\ \log \pi_K \end{pmatrix}$ ← 各成分

$$\bullet \mathbb{E}[\log \pi_k] = \int \log \pi_k C_D(\alpha) \prod_{k=1}^K \pi_k^{\alpha_k-1} d\pi_{1:k-1}$$

第k要素に注目.

$$\begin{aligned} \mathbb{E}[\log \pi_k] &= \int \log \pi_k C_D(\alpha) \prod_{k=1}^K \pi_k^{\alpha_k-1} d\pi_{1:k-1} \\ &= C_D(\alpha) \int \log \pi_k \prod_{k=1}^K \pi_k^{\alpha_k-1} d\pi_{1:k-1} \\ &= C_D(\alpha) \int \pi_1^{\alpha_1-1} \dots \pi_{k-1}^{\alpha_{k-1}-1} \left(\frac{\partial}{\partial \alpha_k} \pi_k^{\alpha_k-1} \right) \pi_{k+1}^{\alpha_{k+1}-1} \dots \pi_K^{\alpha_K-1} d\pi_{1:k-1} \\ &= C_D(\alpha) \frac{\partial}{\partial \alpha_k} \left(\int \prod_{k=1}^K \pi_k^{\alpha_k-1} d\pi_{1:k-1} \right) \\ &= C_D(\alpha) \frac{\partial}{\partial \alpha_k} \left(\frac{1}{C_D(\alpha)} \int C_D(\alpha) \prod_{k=1}^K \pi_k^{\alpha_k-1} d\pi_{1:k-1} \right) \end{aligned}$$

$$\begin{aligned}
&= C_D(\alpha) \frac{\partial}{\partial \alpha_k} \left(\frac{1}{C_D(\alpha)} \int \text{Dir}(\pi | \alpha) d\pi_{1:k-1} \right) \\
&= C_D(\alpha) \frac{\partial}{\partial \alpha_k} \left(\frac{1}{C_D(\alpha)} \right) \\
&= \frac{\partial}{\partial \alpha_k} \log \frac{1}{C_D(\alpha)} \\
&= \frac{\partial}{\partial \alpha_k} \log \frac{\prod_{k'=1}^k \Gamma(\alpha_{k'})}{\Gamma(\sum_{k'=1}^k \alpha_{k'})} \\
&= \frac{\partial}{\partial \alpha_k} \left(\sum_{k'=1}^k \log \Gamma(\alpha_{k'}) - \log \Gamma(\sum_{k'=1}^k \alpha_{k'}) \right) \\
&= \frac{\partial}{\partial \alpha_k} \log \Gamma(\alpha_k) - \frac{\partial}{\partial \alpha_k} \log \Gamma(\sum_{k'=1}^k \alpha_{k'}) \\
&= \psi(\alpha_k) - \psi\left(\sum_{k'=1}^k \alpha_{k'}\right). \\
&\therefore \mathbb{E}[\log \pi_k] = \psi(\alpha_k) - \psi\left(\sum_{k'=1}^k \alpha_{k'}\right)
\end{aligned}$$

• $\mathcal{I} \rightarrow \text{Dir}^0 -$

$$\begin{aligned}
H[\text{Dir}(\pi | \alpha)] &= \mathbb{E}[-\log \text{Dir}(\pi | \alpha)] \\
&= - \sum_{k=1}^k (\alpha_k - 1) \mathbb{E}[\log \pi_k] - \log C_D(\alpha) \\
&= - \sum_{k=1}^k (\alpha_k - 1) \left(\psi(\alpha_k) - \psi\left(\sum_{k'=1}^k \alpha_{k'}\right) \right) - \log C_D(\alpha).
\end{aligned}$$

• KL divergence.

$$p(\pi) = \text{Dir}(\pi | \alpha), \quad q(\pi) = \text{Dir}(\pi | \hat{\alpha}) \text{ 等等.}$$

$$KL[q||p] = -H[\text{Dir}(\pi|\hat{\alpha})] - \mathbb{E}_q[\log \text{Dir}(\pi|\alpha)].$$

$$\begin{aligned} \mathbb{E}_q[\log \text{Dir}(\pi|\alpha)] &= \sum_{k=1}^K (\alpha_k - 1) \mathbb{E}_q[\log \pi_k] + \log C_D(\alpha) \\ &= \sum_{k=1}^K (\alpha_k - 1) (\psi(\hat{\alpha}_k) - \psi(\sum_{k=1}^K \hat{\alpha}_k)) + \log C_D(\alpha) \end{aligned}$$

$$\therefore KL[q||p]$$

$$\begin{aligned} &= \sum_{k=1}^K (\hat{\alpha}_k - 1) (\psi(\hat{\alpha}_k) - \psi(\sum_{k=1}^K \hat{\alpha}_k)) + \log C_D(\hat{\alpha}) \\ &\quad - \sum_{k=1}^K (\alpha_k - 1) (\psi(\hat{\alpha}_k) - \psi(\sum_{k=1}^K \hat{\alpha}_k)) - \log C_D(\alpha) \\ &= \sum_{k=1}^K (\hat{\alpha}_k - \alpha_k) (\psi(\hat{\alpha}_k) - \psi(\sum_{k=1}^K \hat{\alpha}_k)) + \log \frac{C_D(\hat{\alpha})}{C_D(\alpha)}. \end{aligned}$$

* ガンマ分布.

$\lambda \in \mathbb{R}_{>0}$ を生成する分布.

$$\text{pdf. } \text{Gam}(\lambda|a,b) = C_G(a,b) \lambda^{a-1} e^{-b\lambda},$$

$$C_G(a,b) = \frac{b^a}{\Gamma(a)}. \quad (a,b \in \mathbb{R}_{>0}).$$

・ 対数表示.

$$\log \text{Gam}(\lambda|a,b) = (a-1) \log \lambda - b\lambda + \log C_G(a,b).$$

・ 平均.

$$\mathbb{E}[\lambda] = \int_0^{\infty} \lambda \cdot C_G(a,b) \lambda^{a-1} e^{-b\lambda} d\lambda$$

$$\begin{aligned}
&= C_G(a,b) \int_0^{\infty} \lambda^a e^{-b\lambda} d\lambda \\
&= C_G(a,b) \frac{1}{C_G(a+1,b)} \int_0^{\infty} C_G(a+1,b) \lambda^a e^{-b\lambda} d\lambda \\
&= C_G(a,b) \frac{1}{C_G(a+1,b)} \int_0^{\infty} \text{Gam}(\lambda|a+1,b) d\lambda \\
&= \frac{b^a}{\Gamma(a)} = \frac{\cancel{a\Gamma(a)}}{\cancel{\Gamma(a)}} \frac{1}{b} = \frac{a}{b}.
\end{aligned}$$

$$\begin{aligned}
\mathbb{E}[\log \lambda] &= \int_0^{\infty} \log \lambda \cdot C_G(a,b) \lambda^{a-1} e^{-b\lambda} d\lambda \\
&= C_G(a,b) \int_0^{\infty} \left(\frac{\partial}{\partial a} \lambda^{a-1} \right) e^{-b\lambda} d\lambda \\
&= C_G(a,b) \frac{\partial}{\partial a} \int_0^{\infty} \lambda^{a-1} e^{-b\lambda} d\lambda \\
&= C_G(a,b) \frac{\partial}{\partial a} \frac{1}{C_G(a,b)} \int_0^{\infty} \text{Gam}(\lambda|a,b) d\lambda \\
&= C_G(a,b) \frac{\partial}{\partial a} \frac{1}{C_G(a,b)} \quad = 1 \\
&= \frac{\partial}{\partial a} \log \frac{1}{C_G(a,b)} \\
&= \frac{\partial}{\partial a} (\log \Gamma(a) - a \log b) \\
&= \psi(a) - \log b.
\end{aligned}$$

• $\mathcal{I} = \text{IPI}^\circ -$.

$$\begin{aligned}
 & H[\text{Gam}(\lambda|a,b)] \\
 &= -\mathbb{E}[\log \text{Gam}(\lambda|a,b)] \\
 &= -\mathbb{E}[\log C_{\mathcal{G}}(a,b) + (a-1)\log \lambda - b\lambda] \\
 &= -(a-1)\mathbb{E}[\log \lambda] + b\mathbb{E}[\lambda] - \log C_{\mathcal{G}}(a,b) \\
 &= (1-a)\psi(a) - (1-a)\log b + b \cdot \frac{a}{b} - a \log b + \log \Gamma(a) \\
 &= (1-a)\psi(a) - \log b + a + \log \Gamma(a).
 \end{aligned}$$

• $p(\lambda) = \text{Gam}(\lambda|a,b)$ & $q(\lambda) = \text{Gam}(\lambda|\hat{a},\hat{b})$ \Rightarrow KL-divergence.

$$\text{KL}[q(\lambda)||p(\lambda)] = -H[\text{Gam}(\lambda|\hat{a},\hat{b})] - \mathbb{E}_q[\log \text{Gam}(\lambda|a,b)].$$

$$\begin{aligned}
 & \mathbb{E}_q[\log \text{Gam}(\lambda|a,b)] \\
 &= (a-1)\mathbb{E}_q[\log \lambda] - b\mathbb{E}_q[\lambda] + \log C_{\mathcal{G}}(a,b) \\
 &= (a-1)(\psi(\hat{a}) - \log \hat{b}) - \frac{b\hat{a}}{\hat{b}} + a(\log b - \log \Gamma(a)).
 \end{aligned}$$

$$\begin{aligned}
 \therefore \text{KL}[q(\lambda)||p(\lambda)] & \\
 &= -(1-\hat{a})\psi(\hat{a}) + \log \hat{b} - \hat{a} - \log \Gamma(\hat{a}) \\
 &\quad - (a-1)(\psi(\hat{a}) - \log \hat{b}) + \frac{b\hat{a}}{\hat{b}} - a(\log b + \log \Gamma(a)). \\
 &= (\hat{a}-a)\psi(\hat{a}) - a \log \frac{b}{\hat{b}} + \hat{a} \left(\frac{b}{\hat{b}} - 1 \right) + \log \frac{\Gamma(a)}{\Gamma(\hat{a})}.
 \end{aligned}$$

* 1次元 Gauss 分布.

$x \in \mathbb{R}$ を生成.

$$\text{pdf. } \mathcal{N}(x|\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right).$$

$\mu \in \mathbb{R}$: 平均パラメータ, $\sigma^2 \in \mathbb{R}_{>0}$: 分散パラメータ.

・ 対数表示.

$$\begin{aligned} & \log \mathcal{N}(x|\mu, \sigma^2) \\ &= -\frac{1}{2} (\log 2\pi + \log \sigma^2) + \left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) \\ &= -\frac{1}{2} \left(\frac{1}{\sigma^2}(x-\mu)^2 + \log \sigma^2 + \log 2\pi \right). \end{aligned} \quad \leftarrow x \text{ に関する凸の二次関数}$$

・ 平均.

$$\begin{aligned} \mathbb{E}[X] &= \int_{-\infty}^{\infty} x \mathcal{N}(x|\mu, \sigma^2) dx \\ &= \int_{-\infty}^{\infty} (x-\mu) \mathcal{N}(x|\mu, \sigma^2) dx + \underbrace{\mu \int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) dx}_{=1} \\ &= \mu + \int_{-\infty}^{\infty} (x-\mu) \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx \\ &= \mu + \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{x-\mu}{\sigma} \exp\left(-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2\right) dx \\ &= \mu + \frac{\sigma}{\sqrt{2\pi}} \int_{-\infty}^{\infty} z e^{-\frac{z^2}{2}} dz. \end{aligned} \quad \downarrow z = \frac{x-\mu}{\sigma}$$

$$\begin{aligned}
&= \mu - \frac{\sigma}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left(\frac{d}{dz} e^{-\frac{z^2}{2}} \right) dz \\
&= \mu - \frac{\sigma}{\sqrt{2\pi}} \left[e^{-\frac{z^2}{2}} \right]_{-\infty}^{\infty} \\
&= \mu.
\end{aligned}$$

(別の計算方法)

$$\begin{aligned}
\mathbb{E}[X] &= \mu + \int_{-\infty}^{\infty} (x - \mu) \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx \\
&= \mu + \sigma^2 \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \frac{\partial}{\partial \mu} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx \\
&= \mu + \sigma^2 \frac{\partial}{\partial \mu} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx \\
&\qquad\qquad\qquad = \mathcal{N}(x | \mu, \sigma^2) \\
&= \mu + \sigma^2 \frac{\partial}{\partial \mu} 1 \\
&= \mu.
\end{aligned}$$

$$\begin{aligned}
 \mathbb{E}[X^2] &= \int_{-\infty}^{\infty} x^2 \mathcal{N}(x|\mu, \sigma^2) dx \\
 &= \int_{-\infty}^{\infty} (x-\mu)^2 \mathcal{N}(x|\mu, \sigma^2) dx + 2\mu \underbrace{\int_{-\infty}^{\infty} x \mathcal{N}(x|\mu, \sigma^2) dx}_{=\mathbb{E}[X]=\mu} \\
 &\quad - \underbrace{\mu^2 \int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) dx}_{=1}
 \end{aligned}$$

$$= \mu^2 + \int_{-\infty}^{\infty} (x-\mu)^2 \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx$$

$$= \mu^2 + \frac{\sigma}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left(\frac{x-\mu}{\sigma}\right)^2 \exp\left(-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2\right) dx$$

$$= \mu^2 + \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \underbrace{z^2 e^{-\frac{z^2}{2}} dz}_{\text{偶函数}} \quad \swarrow z = \frac{x-\mu}{\sigma}$$

$$= \mu^2 + \frac{\sigma^2}{\sqrt{2\pi}} 2 \int_0^{\infty} z^2 e^{-\frac{z^2}{2}} dz$$

$$= \mu^2 - \sigma^2 \sqrt{\frac{2}{\pi}} \int_0^{\infty} z \left(\frac{d}{dz} e^{-\frac{z^2}{2}}\right) dz$$

$$= \mu^2 - \sigma^2 \sqrt{\frac{2}{\pi}} \left(\left[z e^{-\frac{z^2}{2}} \right]_0^{\infty} - \int_0^{\infty} e^{-\frac{z^2}{2}} dz \right)$$

$$= \mu^2 + \sigma^2 \sqrt{\frac{2}{\pi}} \int_0^{\infty} e^{-\frac{z^2}{2}} dz$$

↳ Gauss積分

$$= \mu^2 + \sigma^2 \sqrt{\frac{2}{\pi}} \cdot \frac{1}{2} \sqrt{2\pi} = \mu^2 + \sigma^2.$$

$$\therefore V[X] = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \sigma^2.$$

$$\begin{aligned}
&= \frac{1}{2} \left(\frac{1}{\sigma^2} \mathbb{E}[(x-\mu)^2] + \log \sigma^2 + \log 2\pi \right) \\
&= \frac{1}{2} \left(\frac{1}{\sigma^2} V[x] + \log \sigma^2 + \log 2\pi \right) \\
&= \frac{1}{2} \left(1 + \log \sigma^2 + \log 2\pi \right). \quad \text{we } \mu = \text{mean}
\end{aligned}$$

• $p(x) = \mathcal{N}(x | \mu, \sigma^2)$, $q(x) = \mathcal{N}(x | \hat{\mu}, \hat{\sigma}^2)$ or KL-divergence.

$$\text{KL}[q(x) \| p(x)]$$

$$= -H[q(x)] - \mathbb{E}_q[\log p(x)].$$

$$= -\frac{1}{2} \left(1 + \log \hat{\sigma}^2 + \log 2\pi \right) - \mathbb{E}_q[\log p(x)].$$

$$\mathbb{E}_q[\log p(x)]$$

$$= \mathbb{E}_q \left[-\frac{1}{2} \left(\frac{1}{\sigma^2} (x-\mu)^2 + \log \sigma^2 + \log 2\pi \right) \right]$$

$$= -\frac{1}{2} \left(\frac{1}{\sigma^2} \mathbb{E}_q[x^2 - 2\mu x + \mu^2] + \log \sigma^2 + \log 2\pi \right)$$

$$= -\frac{1}{2} \left(\frac{1}{\sigma^2} (\hat{\mu}^2 + \hat{\sigma}^2 - 2\mu\hat{\mu} + \mu^2) + \log \sigma^2 + \log 2\pi \right).$$

$$= -\frac{1}{2} \left(\frac{1}{\sigma^2} ((\mu - \hat{\mu})^2 + \hat{\sigma}^2) + \log \sigma^2 + \log 2\pi \right).$$

$$\therefore \text{KL}[q(x) \| p(x)]$$

$$= -\frac{1}{2} \left(1 + \log \hat{\sigma}^2 + \log 2\pi \right)$$

$$+ \frac{1}{2} \left(\frac{1}{\sigma^2} ((\mu - \hat{\mu})^2 + \hat{\sigma}^2) + \log \sigma^2 + \log 2\pi \right).$$

$$= \frac{1}{2} \left(\frac{1}{\sigma^2} ((\mu - \hat{\mu})^2 + \hat{\sigma}^2) + \log \frac{\sigma^2}{\hat{\sigma}^2} - 1 \right).$$

Remark Gauss積分について.

• $a > 0, b \in \mathbb{R}$ とし,

$$I(a) := \int_{-\infty}^{\infty} e^{-a(x-b)^2} dx = \sqrt{\frac{\pi}{a}}$$

pf. $\sqrt{a}(x-b) = z$ と変数変換すると, $\frac{dx}{dz} = \frac{1}{\sqrt{a}}$ となる.

$$I(a) = \frac{1}{\sqrt{a}} \int_{-\infty}^{\infty} e^{-z^2} dz. \quad I := \int_{-\infty}^{\infty} e^{-z^2} dz = \sqrt{\pi} \text{ と示す.}$$

$$I^2 = \left(\int_{-\infty}^{\infty} e^{-x^2} dx \right)^2 = \left(\int_{-\infty}^{\infty} e^{-x^2} dx \right) \left(\int_{-\infty}^{\infty} e^{-y^2} dy \right)$$

$$= \int_{\mathbb{R}^2} e^{-(x^2+y^2)} dx dy$$

極座標変換 $x = r \cos \theta, y = r \sin \theta$

Jacobian は $r > 0$.

$$= \int_0^{\infty} \int_0^{2\pi} r e^{-r^2} dr d\theta$$

$$\frac{d}{dr} e^{-r^2} = -2r e^{-r^2}$$

$$= 2\pi \left[-\frac{1}{2} e^{-r^2} \right]_0^{\infty} = \pi.$$

$$\therefore I = \sqrt{\pi}.$$



• $\frac{d}{da} I(a) = -\frac{1}{2a} \sqrt{\frac{\pi}{a}}.$ とし,

$$\frac{d}{da} I(a) = \frac{d}{da} \int_{-\infty}^{\infty} e^{-a(x-b)^2} dx = \int_{-\infty}^{\infty} \frac{d}{da} e^{-a(x-b)^2} dx$$

$$= - \int_{-\infty}^{\infty} (x-b)^2 e^{-a(x-b)^2} dx.$$

$$\therefore \int_{-\infty}^{\infty} (x-b)^2 e^{-a(x-b)^2} dx = \frac{1}{2a} \sqrt{\frac{\pi}{a}}.$$