

2.3 連続確率分布.

* ベータ分布.

$\mu \in (0,1)$: 確率を生成.

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

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

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

↓ ベータ函数の定義.

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

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

・ 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.

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

↑ ガンマ函数の性質
 $\Gamma(x+1) = x \Gamma(x).$

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

$\Sigma = 3\pi^2$,

↑ ベータ函数の形にもつていい感じ
 $\log x$ じゃま.

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

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

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

← 微分の定義

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

を用いてもOK. わんぱく式と

これで用意など、

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

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

↓ 微分と積分の交換は
ここで成立しても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、

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

• $I = \int \partial t^o -$

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

* Dirichlet ブラット.

$$\pi = \begin{pmatrix} \pi_1 \\ \vdots \\ \pi_K \end{pmatrix} \quad K\text{-dim. Prob. vec. を生成}.$$

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

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

• Dirichlet ブラットが確率分布であることの確認.

π : prob. vec. たり, $\pi_i \geq 0$ ($i = 1, \dots, K-1$) が決まる.

$$\pi_K = 1 - \sum_{i=1}^{K-1} \pi_i \quad \text{が決まる. さて,}$$

$$D_K := \left\{ \pi \in \mathbb{R}^K \mid \pi_i \geq 0 \quad (i = 1, \dots, K), \quad \sum_{i=1}^K \pi_i \leq 1 \right\} \text{とく}$$

$D_{K-1} \cap D_{\text{Tr}}(\pi|\alpha)$ を積分する. $d\pi_1 \cdots d\pi_{K-1} = d\pi_{1:K-1} t_{\text{Tr}}$.

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

あとは、以下を主張を示せばよい。

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

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

$$= \int_0^\infty \dots \int_0^\infty \prod_{k=1}^K \left(t_k^{\alpha_{k-1}} e^{-t_k} \right) dt_K.$$

∴ "変換変換" $(t_1, \dots, t_K) \rightarrow (\pi_1, \dots, \pi_{K-1}, x)$ を次のようにして:

$$t_i = \pi_i x \quad (i = 1, \dots, K-1), \quad t_K = \left(1 - \sum_{i=1}^{K-1} \pi_i\right) x. \quad -(*)$$

$$1 - \sum_{i=1}^{K-1} \pi_i =: \pi_K \text{ とおく.}$$

積分範囲 $[0, \infty)^K$ がどの辺に変換されるかを考える:

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

$$\pi_i \geq 0 \text{ and } 1 - \sum_{i=1}^{K-1} \pi_i \geq 0 \text{ and } x \geq 0 \text{ これが条件になります.}$$

$$\begin{cases} \pi_i \leq 0 \text{ and } 1 - \sum_{i=1}^{K-1} \pi_i \leq 0 \text{ and } x \leq 0 \text{ も考慮すればよい}, \\ \text{改めて } \pi'_i = -\pi_i, \quad 1 - \sum_{i=1}^{K-1} \pi'_i = -\left(1 - \sum_{i=1}^{K-1} \pi_i\right), \quad x' = -x \end{cases}$$

と変数変換すればこの Jacobian の絶対値は 1 となるOK.

ゆえに、積分範囲は、 $\mathcal{S} := \underbrace{D_{K-1}}_{\pi_i > 0} \times \underbrace{[0, \infty)}_x$ となります.

また、変数変換 (*) の Jacobian は 1 のとおり.

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

$$= \det \left(\begin{array}{cccc|c} x & & 0 & \pi_1 \\ & x & & \pi_2 \\ 0 & & \ddots & \vdots \\ \hline -x & -x & \cdots & -x & \pi_K \end{array} \right) = x^{K-1} (\geq 0).$$

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

$$= \int_D \prod_{k=1}^K \left((\pi_k x)^{\alpha_k-1} e^{-\pi_k x} \right) x^{K-1} d\pi_{1:K-1} dx$$

↑ Jacobian の計算

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

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



• 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.

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

第 k 要素 := 注目.

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

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

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

$$\frac{\prod_{f=1}^K \Gamma(\sum_{f=1}^K \alpha_f)}{\prod_{f=1}^K \Gamma(\alpha_f)} = \frac{\alpha_k}{\sum_{f=1}^K \alpha_f}$$

$$\therefore \mathbb{E}[\pi_k] = \frac{1}{\sum_{f=1}^K \alpha_f} \alpha_k. \quad \log \pi_k = \begin{pmatrix} \log \pi_1 \\ \vdots \\ \log \pi_K \end{pmatrix} \text{を略記.}$$

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

第 k 要素に注目.

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

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

$$\therefore \mathbb{E}[\log \pi] = \varphi(\alpha) - \varphi\left(\sum_{k=1}^K \alpha_k\right)$$

• I>t口t° - .

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

• KL divergence.

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

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

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

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

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

* ガンマ分布.

$\lambda \in \mathbb{R}_{>0}$ の生成分布.

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

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

• 对数表示.

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

• 平均.

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

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

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

- エントロピー.

$$H[\text{Gam}(\lambda|a,b)]$$

$$= -\mathbb{E}[\log \text{Gam}(\lambda|a,b)]$$

$$= -\mathbb{E}[\log C_F(a,b) + (a-1)\log \lambda - b\lambda]$$

$$= -(a-1)\mathbb{E}[\log \lambda] + b\mathbb{E}[\lambda] - \log C_F(a,b)$$

$$= (1-a)\psi(a) - (1-a)\log b + b \cdot \frac{a}{b} - a \log b + \log \Gamma(a)$$

$$= (1-a)\psi(a) - \log b + a + \log \Gamma(a).$$

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

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

$$\mathbb{E}_q[\log \text{Gam}(\lambda|a,b)]$$

$$= (a-1)\mathbb{E}_q[\log \lambda] - b\mathbb{E}_q[\lambda] + \log C_F(a,b)$$

$$= (a-1)(\psi(\hat{a}) - \log \hat{b}) - \frac{b\hat{a}}{\hat{b}} + a(\log b - \log \Gamma(a)).$$

$$\therefore \text{KL}[q_f(\lambda) \| p(\lambda)]$$

$$= -(1-\hat{a})\psi(\hat{a}) + \log \hat{b} - \hat{a} - \log \Gamma(\hat{a})$$

$$- (a-1)(\psi(\hat{a}) - \log \hat{b}) + \frac{b\hat{a}}{\hat{b}} - a(\log b + \log \Gamma(a)).$$

$$= (\hat{a}-a)\psi(\hat{a}) - a \log \frac{b}{\hat{b}} + \hat{a} \left(\frac{b}{\hat{b}} - 1 \right) + \log \frac{\Gamma(a)}{\Gamma(\hat{a})}.$$

* 1次元 Gauss 分布.

$x \in \mathbb{R}$ を生成.

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

$\mu \in \mathbb{R}$: 平均 / メンツ, $\sigma^2 \in \mathbb{R}_{>0}$: 分散 / メンツ.

・ 对数表示.

$$\begin{aligned} & \log \mathcal{N}(x|\mu, \sigma^2) \\ &= -\frac{1}{2}(\log 2\pi + \log \sigma^2) + \left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) \\ &= -\frac{1}{2}\left(\frac{1}{\sigma^2}(x-\mu)^2 + \log \sigma^2 + \log 2\pi\right). \end{aligned} \quad \leftarrow \text{右辺は凸関数}$$

・ 平均.

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x \mathcal{N}(x|\mu, \sigma^2) dx$$

$$= \int_{-\infty}^{\infty} (x-\mu) \mathcal{N}(x|\mu, \sigma^2) dx + \mu \underbrace{\int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) dx}_{=1}$$

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

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

$$= \mu + \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} z e^{-\frac{z^2}{2}} dz. \quad \downarrow z = \frac{x-\mu}{\sigma}$$

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

(別の計算方法)

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

$$\begin{aligned}
\mathbb{E}[X^2] &= \int_{-\infty}^{\infty} x^2 N(x|\mu, \sigma^2) dx \\
&= \int_{-\infty}^{\infty} (x-\mu)^2 N(x|\mu, \sigma^2) dx + 2\mu \int_{-\infty}^{\infty} x N(x|\mu, \sigma^2) dx \\
&\quad - \mu^2 \int_{-\infty}^{\infty} N(x|\mu, \sigma^2) dx \\
&\quad = \underline{\underline{\mathbb{E}[X] = \mu}} \\
&= \mu^2 + \int_{-\infty}^{\infty} (x-\mu)^2 \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx \\
&= \mu^2 + \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left(\frac{x-\mu}{\sigma}\right)^2 \exp\left(-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2\right) dx \\
&= \mu^2 + \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{z^2 e^{-\frac{z^2}{2}}}{\sqrt{2\pi}} dz \quad \downarrow z = \frac{x-\mu}{\sigma} \\
&\quad \text{偶函数} \\
&= \mu^2 + \frac{\sigma^2}{\sqrt{2\pi}} 2 \int_0^{\infty} z^2 e^{-\frac{z^2}{2}} dz \\
&= \mu^2 - \sigma^2 \sqrt{\frac{2}{\pi}} \int_0^{\infty} z \left(\frac{d}{dz} e^{-\frac{z^2}{2}} \right) dz \\
&= \mu^2 - \sigma^2 \sqrt{\frac{2}{\pi}} \left(\left[ze^{-\frac{z^2}{2}} \right]_0^\infty - \int_0^\infty e^{-\frac{z^2}{2}} dz \right) \\
&= \mu^2 + \sigma^2 \sqrt{\frac{2}{\pi}} \int_0^\infty e^{-\frac{z^2}{2}} dz \\
&\quad \text{Gauss 標準分布} \\
&= \mu^2 + \sigma^2 \sqrt{\frac{2}{\pi}} \cdot \frac{1}{2} \sqrt{2\pi} = \mu^2 + \sigma^2. \\
\therefore \mathbb{V}[X] &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \sigma^2.
\end{aligned}$$

(別の計算方法) ($\sigma^2 = \lambda^{-1}$ とも書く。)

$$\begin{aligned}
 \mathbb{E}[X^2] &= \mu^2 + \int_{-\infty}^{\infty} (x-\mu)^2 \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx \\
 &= \mu^2 + \int_{-\infty}^{\infty} (x-\mu)^2 \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2}\lambda(x-\mu)^2\right) dx \\
 &= \mu^2 - \frac{2}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} \frac{\partial}{\partial\lambda} \exp\left(-\frac{1}{2}\lambda(x-\mu)^2\right) dx \\
 &= \mu^2 - \frac{2}{\sqrt{2\pi\sigma^2}} \frac{\partial}{\partial\lambda} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2}\lambda(x-\mu)^2\right) dx \\
 &= \mu^2 - \frac{2}{\cancel{\sqrt{2\pi\sigma^2}}} \frac{\partial}{\partial\lambda} \cancel{\sqrt{\frac{2\pi}{\lambda}}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx \\
 &\quad = \mathcal{N}(x|\mu, \sigma^2) \\
 &= \mu^2 - \frac{2}{\sqrt{\sigma^2}} \cdot \left(-\frac{1}{2} \lambda^{-\frac{3}{2}}\right) \\
 &= \mu^2 + \frac{1}{\sqrt{\sigma^2}} \cdot \sigma^2 \cancel{\sqrt{\sigma^2}} \\
 &= \mu^2 + \sigma^2.
 \end{aligned}$$

・ イントロダクション。

$$\begin{aligned}
 H[\mathcal{N}(x|\mu, \sigma^2)] &= -\mathbb{E}[\log \mathcal{N}(x|\mu, \sigma^2)] \\
 &= -\mathbb{E}\left[-\frac{1}{2} \left(\frac{1}{\sigma^2}(x-\mu)^2 + \log \sigma^2 + \log 2\pi \right)\right]
 \end{aligned}$$

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

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

$$KL[q(x) \| p(x)]$$

$$\begin{aligned}
&= -H[q(x)] - \mathbb{E}_q[\log p(x)] \\
&= -\frac{1}{2} \left(1 + \log \hat{\sigma}^2 + \log 2\pi \right) - \mathbb{E}_q[\log p(x)]
\end{aligned}$$

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

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

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

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

Remark Gauss積分について.

- $a > 0, b \in \mathbb{R}$ とする.

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

pf. $\sqrt{a}(x-b) = z \Leftrightarrow$ 座標変換すると, $\frac{dx}{dz} = \frac{1}{\sqrt{a}}$ つまり z^2 ,

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

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

$$= \int_{\mathbb{R}^2} e^{-(x^2+y^2)} dx dy \quad \begin{matrix} \text{極座標変換} \\ x = r \cos \theta, y = r \sin \theta \end{matrix}$$

\downarrow Jacobian は $r > 0$.

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

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

- $\frac{d}{da} I(a) = -\frac{1}{2a} \sqrt{\frac{\pi}{a}}.$ \Leftrightarrow 3式.

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

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