

1 問題設定

2 ベイズ推論

2.1 a と b の分布

2.2 ノイズ σ の推定

前のファイルを参照

2.3 モデル選択

$$y = ax + b \quad \text{か} \quad y = ax \quad \text{か}$$

たとえば …

フックの法則

$$f = kx + \cancel{f_0}$$

ニュートンの運動方程式

$$f = ma + \cancel{f_0}$$

など $y = ax$ 型の理論が選ばれることもある

$y = ax$ の理論を作る

$$\mathbb{X} = \{x_1, x_2, \dots, x_N\}, \mathbb{Y} = \{y_1, y_2, \dots, y_N\}$$

$$y = ax$$

$$a_0 = \frac{\overline{xy}}{\overline{x^2}} \quad (\text{MAP 推定})$$

$$p(a | \mathbb{X}, \mathbb{Y}) = \sqrt{\frac{Nx^2}{2\pi\sigma^2}} \exp\left(-\frac{Nx^2}{2\sigma^2}(a - a_0)^2\right)$$

$p(\sigma | \mathbb{X}, \mathbb{Y})$ はどうなるか？

$$\begin{aligned} E(a) &= \frac{1}{N} \sum_{i=1}^N (y_i - ax_i)^2 \\ &= \frac{1}{N} \sum_{i=1}^N \{y_i^2 - 2ax_iy_i + a^2x_i^2\} \\ &= \frac{1}{N} \sum_{i=1}^N y_i^2 - 2a \frac{1}{N} \sum_{i=1}^N x_iy_i + a^2 \frac{1}{N} \sum_{i=1}^N x_i^2 \\ &= \overline{y^2} - 2a\overline{xy} + a^2\overline{x^2} \\ \frac{\partial E(a)}{\partial a} &= -2\overline{xy} + 2a\overline{x^2} = 0 \end{aligned}$$

よって

$$a = \frac{\overline{xy}}{\overline{x^2}}$$

のとき $E(a)$ が最小。平方完成すると

$$\begin{aligned} E(a) &= \overline{x^2} \left(a^2 - 2a \frac{\overline{xy}}{\overline{x^2}} + \left(\frac{\overline{xy}}{\overline{x^2}} \right)^2 - \left(\frac{\overline{xy}}{\overline{x^2}} \right)^2 \right) + y^2 \\ &= \overline{x^2} \left(a - \frac{\overline{xy}}{\overline{x^2}} \right)^2 + \frac{(\overline{xy})^2}{\overline{x^2}} + \overline{y^2} \\ a_0 &= \frac{\overline{xy}}{\overline{x^2}} \end{aligned}$$

$$E(a_0) = \frac{(\overline{xy})^2}{\overline{x^2}} + \overline{y^2}$$

したがって $E(a)$ は次のように書き換えられる

$$E(a) = E(a_0) + \overline{x^2}(a - a_0)^2$$

a の分布

$$y_i = ax_i + n_i$$

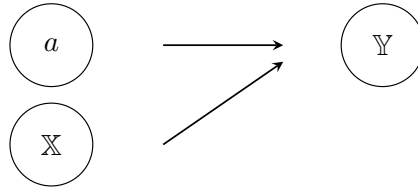
$$n_i \sim N(0, \sigma^2)$$

$$p(n_i) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{n_i^2}{2\sigma^2}\right)$$

$$\mathbb{X} = \{x_1, x_2, \dots, x_N\}$$

$$\mathbb{Y} = \{y_1, y_2, \dots, y_N\}$$

$p(a | \mathbb{X}, \mathbb{Y})$ が知りたい



同時分布

$$p(\mathbb{X}, \mathbb{Y}, a)$$

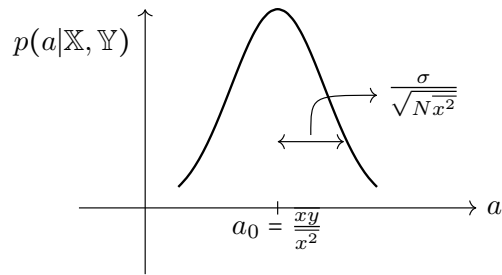
$$\begin{aligned}
 p(\mathbb{X}, \mathbb{Y}, a) &= p(\mathbb{Y} | \mathbb{X}, a)p(\mathbb{X})p(a) \\
 p(\mathbb{Y} | \mathbb{X}, a) &= \prod_{i=1}^N \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y_i - ax_i)^2}{2\sigma^2}\right) \\
 &= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} \frac{1}{N} \sum_{i=1}^N (y_i - a_i)^2\right) \\
 &= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a)\right) \\
 E(a) &= E(a_0) + \overline{x^2}(a - a_0)^2 \\
 a_0 &= \frac{\overline{xy}}{\overline{x^2}}
 \end{aligned}$$

$$\begin{aligned}
 p(\mathbb{X}, \mathbb{Y}, a) &= p(\mathbb{Y} | \mathbb{X}, a)p(\mathbb{X})p(a) \\
 &= p(a | \mathbb{X}, \mathbb{Y})p(\mathbb{X}, \mathbb{Y}) \\
 \therefore p(a | \mathbb{X}, \mathbb{Y}) &\propto p(\mathbb{Y} | \mathbb{X}, a) \\
 &= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a)\right) \\
 &= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0)\right) \exp\left(-\frac{N\overline{x^2}}{2\sigma^2} (a - a_0)^2\right) \\
 &\propto \exp\left(-\frac{N\overline{x^2}}{2\sigma^2} (a - a_0)^2\right)
 \end{aligned}$$

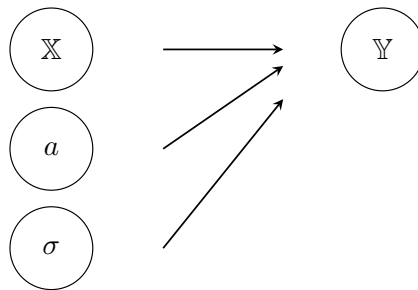
$\int da p(a | \mathbb{X}, \mathbb{Y}) = 1$ より

$$p(a | \mathbb{X}, \mathbb{Y}) = \sqrt{\frac{N\overline{x^2}}{2\pi\sigma^2}} \exp\left(-\frac{N\overline{x^2}}{2\sigma^2} (a - a_0)^2\right)$$

となる



ノイズ σ の推定



生成モデル

$$p(\mathbb{Y} | \mathbb{X}, a, \sigma)$$

$$p(\sigma | \mathbb{X}, \mathbb{Y}) \rightarrow \sigma \text{ を決めて}$$

$$p(a | \mathbb{X}, \mathbb{Y}, \sigma)$$

が得られる

$p(\mathbb{X}, \mathbb{Y}, a, \sigma)$ を考える

$$p(\mathbb{X}, \mathbb{Y}, \sigma) = \int da \underbrace{p(\mathbb{X}, \mathbb{Y}, a, \sigma)}_{p(\sigma)p(a)p(\mathbb{X})p(\mathbb{Y}|\mathbb{X}, a, \sigma)}$$

$$\propto \int da p(\sigma)p(a)p(\mathbb{X})p(\mathbb{Y} | \mathbb{X}, a, \sigma),$$

$$p(\mathbb{X}, \mathbb{Y}, \sigma) = p(\sigma | \mathbb{X}, \mathbb{Y})p(\mathbb{X}, \mathbb{Y})$$

$$\therefore p(\sigma | \mathbb{X}, \mathbb{Y}) \propto \int da p(\mathbb{Y} | \mathbb{X}, a, \sigma)$$

$$= \int da (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a)\right)$$

$$= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0)\right) \int da \exp\left(-\frac{Nx^2}{2\sigma^2} (a - a_0)^2\right)$$

$$= (2\pi\sigma^2)^{-\frac{N-1}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0)\right) \times \sqrt{\frac{1}{Nx^2}}$$

MAP 推定

$$\begin{aligned} F(\sigma) &= -\log p(\sigma | \mathbb{X}, \mathbb{Y}) \\ &= \frac{N}{2\sigma^2} E(a_0) + \frac{N-1}{2} \log(2\pi\sigma^2) \\ \text{変数変換 } S &= \sigma^2 \\ F(S) &= \frac{N}{2S} E(a_0) + \frac{N-1}{2} \log(2\pi S) \\ \frac{dF(S)}{dS} &= -\frac{N}{2S^2} E(a_0) + \frac{N-1}{2S} = 0 \end{aligned}$$

S の微分 = 0 より

$$\begin{aligned} S(N-1) &= NE(a_0) \\ S &= \frac{N}{N-1} E(a_0) \\ \sigma^2 &= \frac{N}{N-1} \frac{1}{N} \sum_{i=1}^N (y_i - a_0 x_i)^2 \\ &= \frac{1}{N-1} \sum_{i=1}^N (y_i - a_0 x_i)^2 \end{aligned}$$

$y = ax$ の理論を作る

$$\mathbb{X} = \{x_1, x_2, \dots, x_N\}, \mathbb{Y} = \{y_1, y_2, \dots, y_N\}$$

$$y = ax$$

$$a_0 = \frac{\overline{xy}}{\overline{x^2}} \quad (\text{MAP 推定})$$

$$p(a | \mathbb{X}, \mathbb{Y}) = \sqrt{\frac{Nx^2}{2\pi\sigma^2}} \exp\left(-\frac{Nx^2}{2\sigma^2} (a - a_0)^2\right)$$

$$p(\sigma | \mathbb{X}, \mathbb{Y}) ? ?$$

σ は

$$\begin{aligned} \sigma^2 &= \frac{N}{N-1} E(a_0) \\ &= \frac{1}{N-1} \sum_{i=1}^N (y_i - a_0 x_i)^2 \end{aligned}$$

と書ける

モデル選択

$$\begin{aligned} K = 1 & \quad y = ax \\ K = 2 & \quad y = ax + b \end{aligned}$$

同時分布

$$\begin{aligned} p(\mathbb{X}, \mathbb{Y}, a, b, K = 2) \\ p(\mathbb{X}, \mathbb{Y}, a, K = 1) \end{aligned}$$

$K = 2$ を考える

$$\begin{aligned} & p(\mathbb{X}, \mathbb{Y}, a, b, K = 2) \\ &= p(K = 2)p(a, b | K = 2)p(\mathbb{Y} | \mathbb{X}, a, b)p(\mathbb{X}) \\ & \quad p(\mathbb{X}, \mathbb{Y}, K = 2) \\ &= \int da \int db p(\mathbb{X}, \mathbb{Y}, a, b, K = 2) \quad (\text{周辺化}) \\ &= p(K = 2 | \mathbb{X}, \mathbb{Y})p(\mathbb{X}, \mathbb{Y}) \end{aligned}$$

$$\begin{aligned} \therefore p(K = 2 | \mathbb{X}, \mathbb{Y}) &\propto \int da \int db p(\mathbb{X}, \mathbb{Y}, a, b, K = 2) \\ &\propto \int da \int db p(\mathbb{Y} | \mathbb{X}, a, b) \\ &= \int da \int db (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} \frac{1}{N} \sum_{i=1}^N \{y_i - (ax_i + b)\}^2\right) \\ &= \int da \int db (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a, b)\right) \\ &= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0, b_0)\right) \int da \exp\left(-\frac{N\bar{x}^2}{2\sigma^2} (a - a_0)^2\right) \int db \exp\left(-\frac{N}{2\sigma^2} (b - b_0)^2\right) \\ &= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0, b_0)\right) \sqrt{\frac{2\pi\sigma^2}{N\bar{x}^2}} \times \sqrt{\frac{2\pi\sigma^2}{N}} \end{aligned}$$

$$\begin{aligned} F(K = 2) &= -\log p(K = 2 | \mathbb{X}, \mathbb{Y}) \\ &= N \left\{ \frac{1}{2\sigma^2} E(a_0, b_0) + \frac{\log N}{2N} + \frac{\log N}{2N} + O\left(\frac{1}{N}\right) \right\} \\ &= N \left\{ \frac{1}{2\sigma^2} E(a_0, b_0) + \frac{\log N}{N} + O\left(\frac{1}{N}\right) \right\} \end{aligned}$$

$K = 1$ を考える

$$\begin{aligned}
& p(\mathbb{X}, \mathbb{Y}, a, K = 1) \\
&= p(K = 1) p(a | K = 1) p(\mathbb{Y} | \mathbb{X}, a) p(\mathbb{X}) \\
& p(\mathbb{X}, \mathbb{Y}, K = 1) \\
&= \int da p(\mathbb{X}, \mathbb{Y}, a, K = 1) \\
&= p(K = 1 | \mathbb{X}, \mathbb{Y}) p(\mathbb{X}, \mathbb{Y})
\end{aligned}$$

$$\begin{aligned}
\therefore p(K = 1 | \mathbb{X}, \mathbb{Y}) &\propto \int da p(\mathbb{X}, \mathbb{Y}, a, K = 1) \\
&= \int da (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} \frac{1}{N} \sum_{i=1}^N (y_i - ax_i)^2\right) \\
&= \int da (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a)\right) \\
&= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0)\right) \int da \exp\left(-\frac{Nx^2}{2\sigma^2} (a - a_0)^2\right) \\
&= (2\pi\sigma^2)^{-\frac{N}{2}} \exp\left(-\frac{N}{2\sigma^2} E(a_0)\right) \times \sqrt{\frac{2\pi\sigma^2}{x^2}}
\end{aligned}$$

$$\begin{aligned}
F(K = 1) &= -\log p(K = 1 | \mathbb{X}, \mathbb{Y}) \\
&= N \left\{ \frac{1}{2\sigma^2} E(a_0) + \frac{\log N}{2N} + O\left(\frac{1}{N}\right) \right\}
\end{aligned}$$