

① Bayes rule

$$p(\bar{\theta} | D) = \frac{\text{posterior } p(\bar{\theta}) \text{ prior } p(D | \bar{\theta})}{\text{evidence } p(D)} = \frac{p(\bar{\theta}) p(D | \bar{\theta})}{\int p(\bar{\theta}) p(D | \bar{\theta}) d\bar{\theta}}$$

$$p(D | \bar{\theta}) = \prod_{n=1}^N p(d_n | \bar{\theta})$$

1) $p(D | \bar{\theta}) \xrightarrow{\bar{\theta}} \max$ maximum likelihood $\bar{\theta}_{ML} = \arg \max_{\bar{\theta}} p(D | \bar{\theta})$

$$\log p(D | \bar{\theta}) = \sum_n \log p(d_n | \bar{\theta})$$

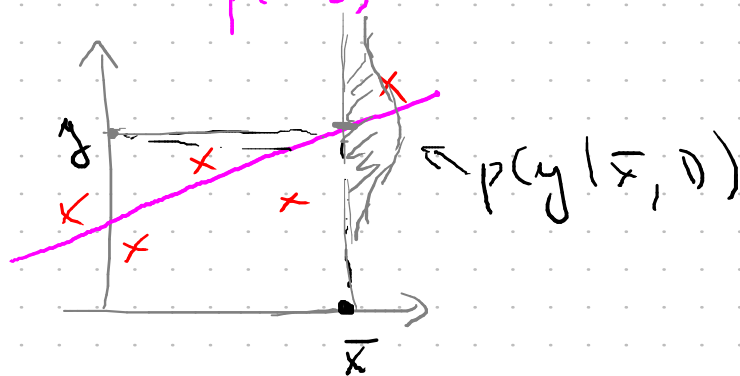
2) $p(\bar{\theta} | D) \xrightarrow{\bar{\theta}} \max$ maximum a posteriori $\bar{\theta}_{MAP} = \arg \max_{\bar{\theta}} p(\bar{\theta} | D)$

$$\log p(\bar{\theta} | D) = \sum_n \log p(d_n | \bar{\theta}) + \log p(\bar{\theta}) + \text{Const}$$

regularizer

3) predictive distribution

$$p(\bar{x} | D) = \int p(\bar{x}, \bar{\theta} | D) d\bar{\theta} = \int \underbrace{p(\bar{x} | \bar{\theta})}_{\text{likelihood}} \underbrace{p(\bar{\theta} | D)}_{\text{posterior}} d\bar{\theta} = \mathbb{E}_{p(\bar{\theta} | D)} [p(\bar{x} | \bar{\theta})]$$



$$p(\text{кробо} | \text{небунуаеи}) = 1\%$$

$$\cancel{p(\text{небун.} | \text{кробо}) = 1\%}$$

$$p(y_{\text{дун}} | \text{дом. пар.}) = 1/2500$$

$$p(y_{\text{дун}} | \text{дом. пар.} \& y_{\text{дун}}) = 7/8 \quad 6/7$$

Bernoulli trials

$$\theta = p(\text{op}^{\text{er}}), \quad D = \text{hththth}$$

$$p(x|\theta) = \theta(1-\theta) \dots = \theta^n \cdot (1-\theta)^m$$

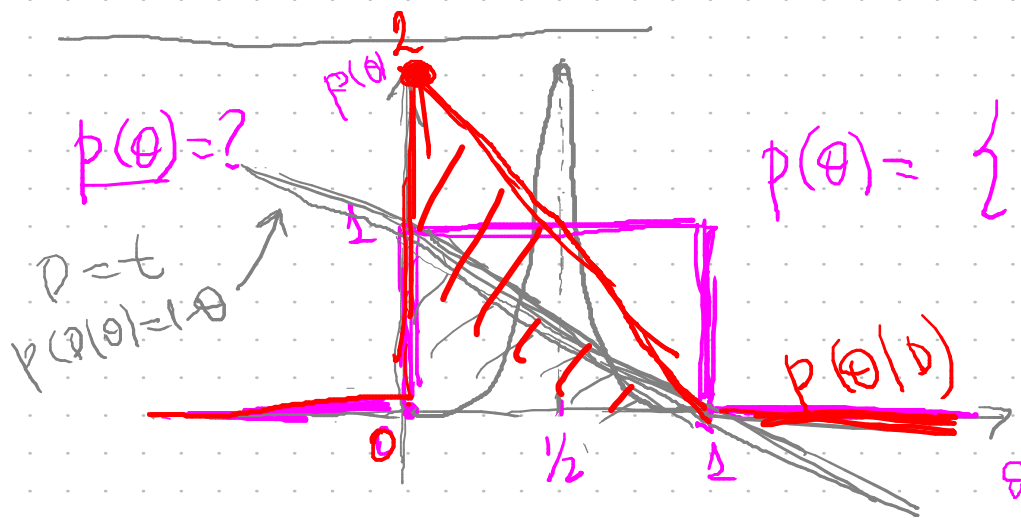
$$p(\theta) \xrightarrow{\theta} \max$$
$$\frac{\partial p(\theta)}{\partial \theta} = n\theta^{n-1}(1-\theta)^m - m\theta^n(1-\theta)^{m-1} = 0$$
$$\theta^{n-1}(1-\theta)^{m-1}(n(1-\theta) - m\theta) = 0$$

$$\theta = \theta_1 \frac{n}{n+m}$$

$$\Theta_{ML} = \frac{n}{n+m}$$

$$D=t \Rightarrow \theta_{ML} = 0$$

$$p(t|\theta) = 1 - \theta$$
$$\theta_{ML} = -\infty$$



$$p(D|\theta) = \theta^n (1-\theta)^m$$

$$p(\theta | D) = \begin{cases} \frac{1}{p(\theta)} \theta^n (1-\theta)^m, & \theta \in [0, 1] \\ 0, & \text{---} \end{cases}$$

$$p(\theta) = \int_0^1 \theta^n (1-\theta)^m d\theta = B(n+1, m+1) =$$

$$= \frac{\Gamma(n+1) \Gamma(m+1)}{\Gamma(\underline{n+m+2})} = \frac{n! \cdot m!}{(n+m+1)!}$$

$$p(\theta|D) = \frac{(n+m+1)!}{n!m!} \theta^n (1-\theta)^m, \quad \theta \in [0,1]$$

↙ θ
max

$$\theta_{\text{MAP}} = \frac{n}{n+m}$$

$$p(\text{open}|D) = \int \overbrace{p(\text{open}|\theta)}^{\theta} \overbrace{p(\theta|D)}^{\theta} d\theta =$$

$$= \int_0^1 \theta \cdot \frac{(n+m+1)!}{n!m!} \theta^n (1-\theta)^m d\theta = \dots \int_0^1 \theta^{n+1} (1-\theta)^m d\theta$$

Bayesian smoothing

$$= \frac{(n+m+1)!}{n!m!} \cdot \frac{(n+1)!m!}{(n+m+2)!}$$

$$p(\text{open}|D) = \frac{n+1}{n+m+2}$$

Laplace's rule of succession