

## Minimal sufficiens

Stikprogefunktionen  $T(y)$  er minimal sufficent for  $\Theta$ , hvis  $\forall y, z \in Y$ :

$$T(y) = T(z) \Leftrightarrow L(\theta; y) \propto L(\theta; z) \text{ for alle } \theta \in \Theta$$

(mod. og tilstr. bet.)

Likelihood derivatens bestemmer en klassedeling af  $Y$ . Kald elementene  $A_y$ .

En stikprogefunktion, der antager en konstant værdi på ethvert  $A_y$  og forskellige værdier på forskellige  $A_y$ 'er, er minimal sufficent.

$$T_1, T_2 \text{ minimal sufficiente} \Rightarrow T_1 = g(T_2),$$

$g$  bijektiv

$$T_1 \text{ min. suff.}, T_2 \text{ suff.} \Rightarrow T_1 = h(T_2)$$

Bemerk

$$\forall y, z \in Y \quad \forall \theta \in \Theta:$$

$$T(y) = T(z) \Leftrightarrow \frac{L(\theta; y)}{L(\theta; z)} \text{ er uafh. af } \theta$$

eks.  $y = (y_1, \dots, y_n)$ ,  $Y_i \sim \text{Cauchy}(1, \theta)$  uafh.

$$L(\theta; y) = \prod_{i=1}^n \frac{1}{\pi} \frac{1}{1 + (y_i - \theta)^2}$$

$(y_{(1)}, y_{(2)}, \dots, y_{(n)})$  er suff. for  $\theta$  (trivialt)

$$\frac{L(\theta; y)}{L(\theta; z)} = \frac{\prod_i (1 + (z_i - \theta)^2)}{\prod_i (1 + (y_i - \theta)^2)}$$

Begge polynomier har koefficienten 1 til  $\theta^{2n}$ , kvot. derfor kan værd. af  $\theta$ , når pol. er ens, desv. når  $(z_1, \dots, z_n)$  er en permutation af  $(y_1, \dots, y_n)$ .

$(y_{(1)}, y_{(2)}, \dots, y_{(n)})$  er altså min. suff.

□

### Eksponentielle familier

$\mathcal{F} = \{f(\cdot; \theta) \mid \theta \in \Theta\}$  udgør en eksponentiel familie, når

$$f(y; \theta) = q(y) \exp \left( \sum_i \gamma_i(\theta) t_i(y) - \tilde{c}(\theta) \right) *$$

værd. af  $\theta$       /      værd. af  $y$       værd. af  $\theta$       værd. af  $y$

Tilhørende likelihood på eksponentiel form.

Bemerk  $f(y; \theta) = 0 \Leftrightarrow q(y) = 0$ , desv. alle tætheder i en eksponentiel familie har samme støtte.

eks.  $Y \sim b(n, \theta)$

$$\begin{aligned}
 f(y; \theta) &= \binom{n}{y} \theta^y (1-\theta)^{n-y}, \quad y = 0, 1, \dots, n \\
 &= \binom{n}{y} \exp \left( y \ln \frac{\theta}{1-\theta} - (-n \ln(1-\theta)) \right)
 \end{aligned}$$

□

eks.  $y = (y_1, y_2, \dots, y_n)$ ,  $Y_i \sim N(\mu, \sigma^2)$  uafh.

$$\begin{aligned} f(y; \theta) &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(y_i-\mu)^2}{2\sigma^2}} \\ &= (2\pi)^{-\frac{n}{2}} \sigma^{-n} \exp\left(-\frac{1}{2\sigma^2}(\sum_i y_i^2 - 2\mu \sum_i y_i + n\mu^2)\right) \\ &= (2\pi)^{-\frac{n}{2}} \exp\left(\frac{\mu}{\sigma^2} \sum_i y_i - \frac{1}{2\sigma^2} \sum_i y_i^2 - \left(\frac{n\mu^2}{2\sigma^2} + \frac{n}{2} \ln \sigma^2\right)\right) \end{aligned}$$

eks.  $y = (y_1, y_2, \dots, y_n)$ ,  $Y_i \sim P(x_i; \theta)$  uafh.,  $x_i$  erne grise

$$\begin{aligned} f(y; \theta) &= \prod_{i=1}^n \frac{e^{-x_i \theta} (x_i \theta)^{y_i}}{y_i!} = \frac{e^{-\theta \sum x_i} \theta^{\sum y_i} \prod x_i^{y_i}}{\prod y_i!} \\ &= \left(\prod_i \frac{x_i^{y_i}}{y_i!}\right) \exp(\ln \theta \sum y_i - \theta \sum x_i) \end{aligned}$$

Når  $\psi_1(\theta), \psi_2(\theta), \dots, \psi_r(\theta)$  er lin. uafh.,

siger \* at vær på reduceret form, og  
r angiver den eksponentielle families orden

Sætning:

\* er på reduceret form  $\Rightarrow$

$T = (T_1(y), T_2(y), \dots, T_r(y))$  er min. suff. for  $\theta$

bevis

\* viser, at  $T$  er sufficiint (Neyman)

betragt  $z, w \in \mathcal{Y}$ , sat  $T_0(\cdot) = \ln q(\cdot)$

$$\lambda(\theta; z) - \lambda(\theta; w)$$

$$\begin{aligned} &= T_0(z) - T_0(w) + \psi_1(\theta)(T_1(z) - T_1(w)) + \\ &\quad \dots + \psi_r(\theta)(T_r(z) - T_r(w)) \end{aligned}$$

$$\lambda(\theta; z) - \lambda(\theta; w) = 0 \Leftrightarrow T_j(z) = T_j(w), j = 0, 1, \dots, r$$

$$L(\theta; z) \propto L(\theta; w) \Leftrightarrow T_j(z) = T_j(w), j = 1, \dots, r$$

$T$  er altså min. suff. for  $\theta$

ekse. logistisk regression

$$y = (y_1, y_2, \dots, y_n), \quad Y_i \sim b(1, \pi_i) \text{ uafh.}$$

$$\pi_i = \frac{\exp(\alpha + \beta x_i)}{1 + \exp(\alpha + \beta x_i)}, \quad i = 1, \dots, n$$

$$(\text{omv. fkt. } \alpha + \beta x_i = \ln \frac{\pi_i}{1 - \pi_i} = \text{logit } \pi_i)$$

$$(x_i, \pi_i) \text{ ligger på } \hat{\pi}(x) = \frac{\exp(\alpha + \beta x)}{1 + \exp(\alpha + \beta x)}$$

grader kældes logistiske kurver

likelihood

$$\begin{aligned} L(\alpha, \beta; y) &= \prod_{i=1}^n \left( \frac{\exp(\alpha + \beta x_i)}{1 + \exp(\alpha + \beta x_i)} \right)^{y_i} \left( \frac{1}{1 + \exp(\alpha + \beta x_i)} \right)^{1-y_i} \\ &= \prod_{i=1}^n (\exp(\alpha + \beta x_i))^{y_i} (1 + \exp(\alpha + \beta x_i))^{-1} \\ &= \exp \sum_{i=1}^n ((\alpha + \beta x_i) y_i - \ln (1 + \exp(\alpha + \beta x_i))) \\ &= \exp (\alpha \sum_i y_i + \beta \sum_i x_i y_i - \sum_i \ln (1 + \exp(\alpha + \beta x_i))) \end{aligned}$$

$(\sum_i y_i, \sum_i x_i y_i)$  er min. suff. for  $(\alpha, \beta)$

r = 2

□

Gentagne observationer (uafh.)

$$y = (y_1, y_2, \dots, y_n) \quad \swarrow \text{reducent form}$$

$$f(y_j; \theta) = q(y_j) \exp \left( \sum_{i=1}^r \pi_i(\theta) t_i(y_j) - \tau(\theta) \right)$$

$$\Rightarrow f(y; \theta) = \left( \prod_{j=1}^n q(y_j) \right) \exp \left( \sum_{i=1}^r \pi_i(\theta) \sum_{j=1}^n t_i(y_j) - n \tau(\theta) \right)$$

der.  $(\sum_{j=1}^n t_1(y_j), \sum_{j=1}^n t_2(y_j), \dots, \sum_{j=1}^n t_r(y_j))$  er

min. suff. for  $\theta$ , dim. fortsat r uafh. af n

Tilstækketige betingelser (ikke nødvendige)  
for eksponentiel familie ( $\mu$ /varis)

- uafh. og identisk for. variable
- støtte afh. ikke af  $\theta$
- eksistens af ikke-triviel suff. stikpr.-fkt.  
med dim. mindre end  $Y$ 's dim. og  
og dim. uafh. af  $n$

Fleve egenskaber ved eksponentielle familier

- $\tau(\theta)$  vkh. ofte diff., når  $\psi_i^{(0)}$ ,  $i=1, \dots, r$   
er diff.
- $\frac{d^r}{d\theta^r} \int_y g(y) f(y; \theta) dv(y)$   
 $= \int_y \frac{d^r}{d\theta^r} g(y) f(y; \theta) dv(y)$ ,  
når integralerne eksisterer  
( $\mu$ /varis)

Regulære eksponentielle familier

$\mathcal{F} = \{ f(\cdot; \theta) \mid \theta \in \Theta \}$  med  $f$  som \*w  
reguler, når

- $\Theta = \{ \theta \mid \int_y g(y) \exp(\sum_i \psi_i(\theta) t_i(y)) dv(y) < \infty \}$
- $\Theta$  åben delmengde af  $\mathbb{R}^k$
- dim. af  $\Theta$  er lig med dim. af den  
min. suff. stikpr.-fkt.
- aff.  $\theta \mapsto \psi(\theta) = (\psi_1(\theta), \psi_2(\theta), \dots, \psi_r(\theta))$  er  
vigtig
- $\psi_i$ 'erne vkh. ofte diff. relt.  $\theta$ 's kompo-  
nenter

els.  $y = (y_1, y_2, \dots, y_n)$

$$Y_1 \sim N(0, \frac{\sigma^2}{1-\rho^2}), \quad |\rho| < 1$$

$$Y_t = \rho Y_{t-1} + E_t, \quad t=2, \dots, n, \quad E_t \sim N(0, \sigma^2) \text{ unif.}$$

Bemerk., at  $Y_t$  eine er. afd.

$$Y \sim N_n(0, \frac{\sigma^2}{1-\rho^2} I_n), \quad I_n = \begin{bmatrix} 1 & \rho & \cdots & \rho^{n-1} \\ \rho & 1 & & \rho^{n-2} \\ \vdots & \vdots & \ddots & \vdots \\ \rho^{n-1} & \rho^{n-2} & \cdots & 1 \end{bmatrix}$$

(autoregressive process of 1. order)

Alternative specification of variable:

$$Y_1 \sim N(0, \frac{\sigma^2}{1-\rho^2}), \quad |\rho| < 1$$

$$Y_t | (Y_1, Y_2, \dots, Y_{t-1}) \sim N(\rho Y_{t-1}, \sigma^2), \quad t=2, \dots, n$$

$$f(y; \rho, \sigma^2) = \frac{\sqrt{1-\rho^2}}{\sqrt{2\pi}\sigma} e^{-\frac{(1-\rho^2)y_1^2}{2\sigma^2}} \prod_{t=2}^n \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(y_t - \rho y_{t-1})^2}{2\sigma^2}}$$

log likelihood:

$$\begin{aligned} 2l(\rho, \sigma^2; y) &= c + \ln(1-\rho^2) - n \ln \sigma^2 - \frac{1}{\sigma^2} y_1^2 + \frac{\rho^2}{\sigma^2} y_1^2 \\ &\quad - \frac{1}{\sigma^2} \sum_{t=2}^n (y_t^2 - 2\rho y_t y_{t-1} + \rho^2 y_{t-1}^2) \\ &= c + \ln(1-\rho^2) - n \ln \sigma^2 - \frac{1}{\sigma^2} (-\rho^2 y_1^2) \\ &\quad - \frac{1}{\sigma^2} \left( \sum_{t=1}^n y_t^2 - 2\rho \sum_{t=2}^n y_t y_{t-1} + \rho^2 \sum_{t=1}^{n-1} y_t^2 \right) \end{aligned}$$

Ist  $\mu$ -regular  
exponential  
familie, da

$$\dim(d_{00}, d_{01}, d_{11}) = 3$$

$$\text{og } \dim(\rho, \sigma^2) = 2$$

$$= c + \ln(1-\rho^2) - n \ln \sigma^2$$

$$- \frac{1}{\sigma^2} (d_{00} - 2\rho d_{01} + \rho^2 d_{11}),$$

$$\text{wohl } d_{rs} = \sum_{t=s+1}^{n-r} y_t y_{t+r-s}$$

$(d_{00}, d_{01}, d_{11})$  er. min. suff. for  $(\rho, \sigma^2)$  □

Egenskaber for negative eksponentielle  
familier af orden 1

$$f(y; \theta) = g(y) \exp(\varphi(\theta)t(y) - \tau(\theta))$$

Bemerk  $\frac{\partial}{\partial \theta} \int_y f(y; \theta) d\nu(y) = 0$

$$\Rightarrow \int_y \frac{\partial}{\partial \theta} f(y; \theta) d\nu(y) = 0$$

$$\int_y f(y; \theta) (\varphi'(\theta)t(y) - \tau'(\theta)) d\nu(y) = 0$$

$$\varphi'(\theta) \int_y t(y) f(y; \theta) d\nu(y)$$

$$- \tau'(\theta) \int_y f(y; \theta) d\nu(y) = 0$$

$$\varphi'(\theta) E[t(Y)] - \tau'(\theta) = 0$$

$$E[t(Y)] = \frac{\tau'(\theta)}{\varphi'(\theta)}$$

ved to gange diff.

$$\int_y f(y; \theta) (\varphi''(\theta)t(y) - \tau''(\theta))$$

$$+ (\varphi'(\theta)t(y) - \tau'(\theta))^2 d\nu(y) = 0$$

$$\varphi''(\theta) E[t(Y)] - \tau''(\theta)$$

$$+ (\varphi'(\theta))^2 E[(t(Y) - E[t(Y)])^2] = 0$$

$$(\varphi'(\theta))^2 \text{Var}[t(Y)] = \tau''(\theta) - \varphi''(\theta) \frac{\tau'(\theta)}{\varphi'(\theta)}$$

$$\text{Var}[t(Y)] = \frac{\tau''(\theta)\varphi'(\theta) - \varphi''(\theta)\tau'(\theta)}{(\varphi'(\theta))^3}$$

ved yderligere diff. kan højre  
ordens momenter tilsvarende  
bestemmes

## Kanoniske parameter

Ved reparametrisering, så  $\gamma = \gamma(\theta)$  valges som ny parameter, bliver udtrykkene for  $E[t(Y)]$  og  $\text{Var}[t(Y)]$  (og for højre ordns momenter) en del simpelere.

$\gamma$  kaldes den kanoniske parameter

$$h(y; \gamma) = q(y) \exp(\gamma + t(y) - \tau_\gamma(t))$$

rhs.

$$Y \sim b(n, \theta)$$

$$f(y; \theta) = \binom{n}{y} \exp\left(y \ln \frac{\theta}{1-\theta} - (-n \ln(1-\theta))\right)$$

$$\text{set } \gamma = \text{logit } \theta = \ln \frac{\theta}{1-\theta}$$

$$\Leftrightarrow 1-\theta = \frac{1}{1+e^\gamma}$$

$$h(y; \gamma) = \binom{n}{y} \exp\left(y \gamma - n \ln(1+e^\gamma)\right)$$

□