

REGRESSION FUNCTION



$$\begin{aligned} E z_{ik} &= 0 \\ \text{Var } z_{ik} &< \infty \\ z_{ik}, u_{ik} &\leftarrow \text{independent} \end{aligned}$$

$\{(u_{ik}, y_{ik})\}$

$$y_{ik} = \mu(u_{ik}) + z_{ik}$$

$$\begin{aligned} R(u) &= E\{y_k | u_k = u\} = E\{\mu(u_k) + z_k | u_k = u\} = \\ &= E\{\mu(u_k) | u_k = u\} + E\{z_k | u_k = u\} = \\ &= \mu(u) + E z_{ik} = \mu(u) \end{aligned}$$

HOW TO ESTIMATE $\mu(u)$?

p.d.f. of u_{ik} :

$$\mu(u) = \frac{\mu(u) \cdot f(u)}{f(u)} = \frac{g(u)}{f(u)} = \frac{\sum_{i=1}^M b_i \varphi_i(u)}{\sum_{i=1}^M a_i \varphi_i(u)}$$

$$\hat{\mu}(u) = \frac{\sum_{k=1}^N y_k K(\frac{u_k - u}{h})}{\sum_{k=1}^N K(\frac{u_k - u}{h})}$$

$$g = \mu \cdot f$$

$$b_i = E y_k \varphi_i(u_k)$$

$$a_i = E \varphi_i(u_k)$$

$$\{(u_k, y_k)\}_{k=1}^N$$

$$\hat{\mu}(u) = \frac{\sum_{k=1}^N y_{ik} \cdot K(\frac{u_{ik} - u}{h})}{\sum_{k=1}^N K(\frac{u_{ik} - u}{h})}$$

$$\hat{\theta} = \frac{\sum \theta_{ik} w_{ik}}{\sum w_{ik}}$$

$$\hat{\mu}(u) = \frac{\sum_{i=1}^S \hat{b}_i \varphi_i(u)}{\sum_{i=1}^S \hat{a}_i \varphi_i(u)}$$

$$\hat{b}_i = \frac{1}{N} \sum_{k=1}^N y_k \varphi_i(u_k)$$

$$\hat{a}_i = \frac{1}{N} \sum_{k=1}^N \varphi_i(u_k)$$