

## 2.1.6 → Kernel Density Estimation

→ overview → one strategy is to assume pdf form + estimate parameters

→ another strategy is to estimate the pdf directly w/  
kernel density estimation

→ Def → Kernel function,  $K(\cdot)$ , is a pdf w/ 2 parameters

→ 1st observed value  $x_i$  → mean of kernel function's distribution  
→ bandwidth  $h$

→ properties → symmetric about  $x_i$

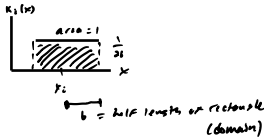
→ notation →  $K_h(\cdot)$  is the kernel function where  $x_i$  is the center of its density  
→ interpretation → bandwidth  $h$  interpretation depends on the  $K(\cdot)$  chosen

→ Basic idea → 1) Choose a density family for  $K(\cdot)$  → Rectangular, triangular, or normal  
→ 2) Estimate  $f(x)$  as the average of  $K_h(x_0), \dots, K_h(x_n)$

→ thus, the kernel density estimate of  $f(x)$  is

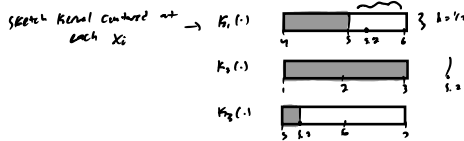
$$\hat{f}(x) = \frac{1}{n} \sum_{i=1}^n K_h(x) \quad (\text{discrete measure})$$

→ Rectangular Kernel →  
(uniform)



→ ex)  $x = (5, 2, 6)$ , use bandwidth  $h=1$  to 1) estimate  $\hat{f}(5.2)$   
2)  $P(X \in 5.2)$

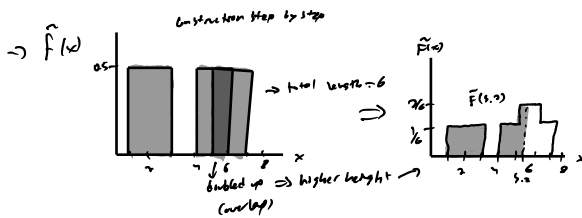
$$\rightarrow 1) \hat{f}(5.2) = \frac{1}{3} [K_1(5.2) + K_1(5.2) + K_1(5.2)]$$



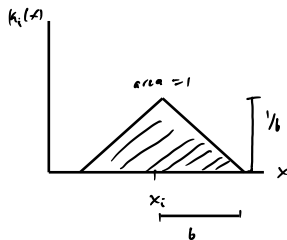
$$\Rightarrow \hat{f}(5.2) = \frac{1}{3} \left( \frac{1}{1} + 0 + \frac{1}{1} \right) = \frac{2}{3} \Rightarrow K_h(x) = \begin{cases} \frac{1}{h} & x_i - h \leq x \leq x_i + h \\ 0 & \text{otherwise} \end{cases}$$

$$\rightarrow 2) \hat{F}(5.2) = \text{area of blue} \Rightarrow \text{average}$$

$$\downarrow = \frac{1}{3} \left( \frac{1.2}{1} + 1 + \frac{0.8}{1} \right) = 0.667$$



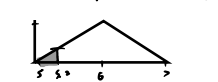
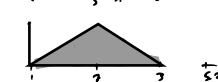
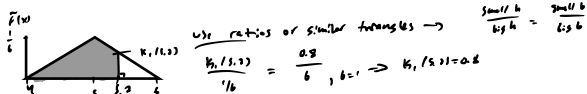
→ Triangular Kernel → assumes isosceles triangle for the kernel function



→ ex)  $x = (5, 2, 6)$ , use bandwidth  $h=1$  to 1) estimate  $\hat{f}(5.2)$   
2)  $P(X \in 5.2)$

$$\rightarrow 1) \hat{f}(5.2) = \frac{1}{3} [K_1(5.2) + K_1(5.2) + K_1(5.2)]$$

$$\downarrow = \frac{1}{3} [0.8 + 0 + 0.2] = \frac{1}{3}$$

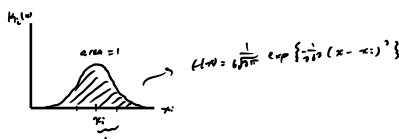


$$\rightarrow 2) \hat{F}(5.2) = \text{area above} \Rightarrow \text{average}$$

$$\downarrow = \frac{1}{3} \left[ \text{area of } (0.8 - \frac{0.8(0.8)}{2}) + 1 + \frac{(0.2)(0.2)}{2} \right]$$

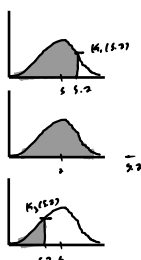
$$\downarrow = 0.667$$

→ Gaussian Kernel → assumes normal density w/ mean  $x_i$  + variance  $h^2 \Rightarrow 1h = h$



→ ex)  $x = (5, 2, 6)$ , use bandwidth  $h=1$  to 1) estimate  $\hat{f}(5.2)$   
2)  $P(X \in 5.2)$

$$\rightarrow 1) \hat{f}(5.2) = \frac{1}{3} [0.391 + 0.002 + 0.290] = 0.228$$



$$\rightarrow 2) \hat{F}(5.2) = \text{area} \Rightarrow \text{average}$$

$$= \frac{1}{3} \left[ \text{norm}(5.2, 5, 1) + \text{norm}(5.2, 2, 1) + \text{norm}(5.2, 6, 1) \right]$$

$$\downarrow = \frac{1}{3} [0.391 + 0.002 + 0.290]$$

$$\downarrow = 0.228$$