

Johns Hopkins Engineering

625.464 Computational Statistics

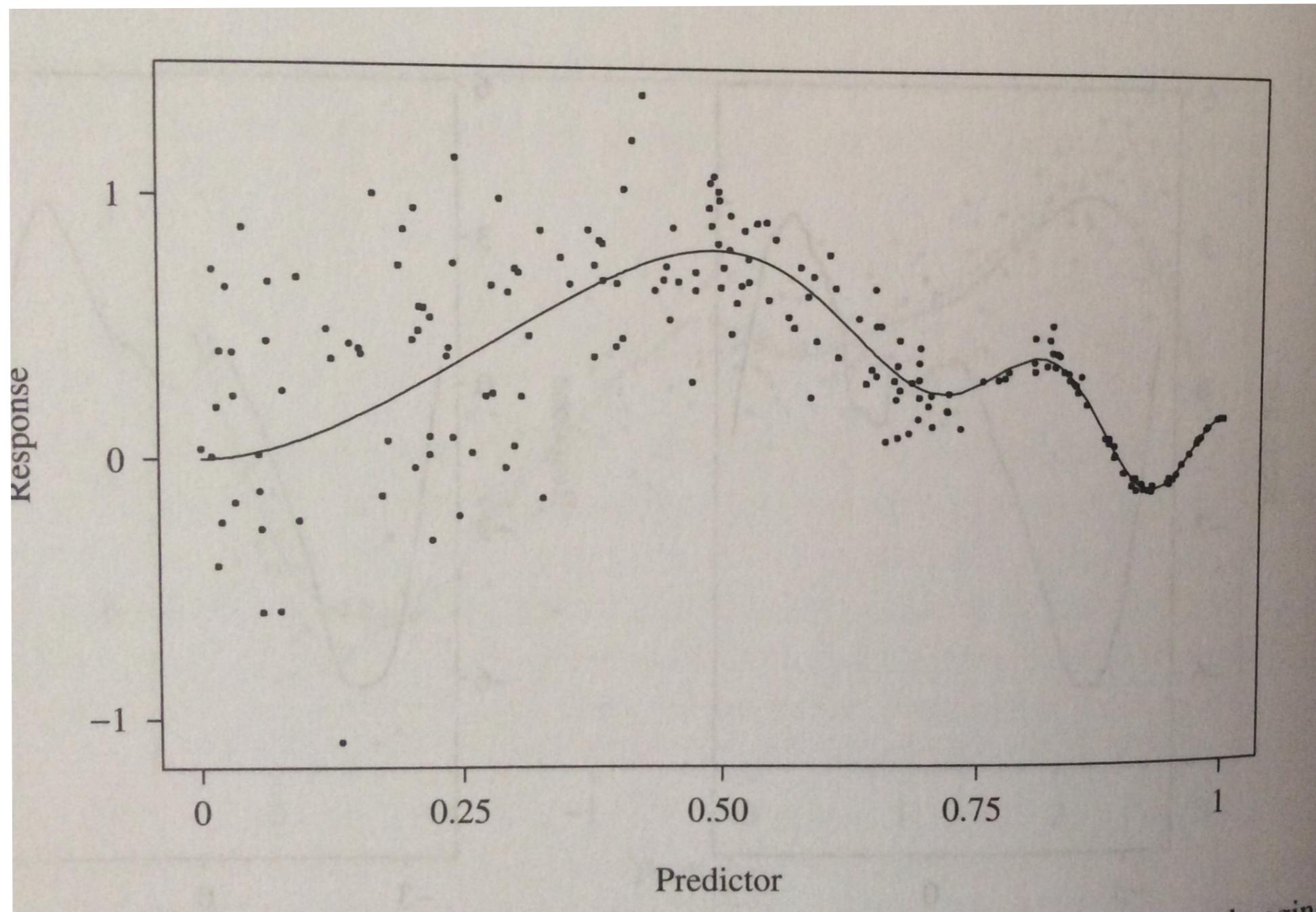
Nonlinear Smoothers

Module 12 Lecture 12D



JOHNS HOPKINS
WHITING SCHOOL
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The Supersmooth



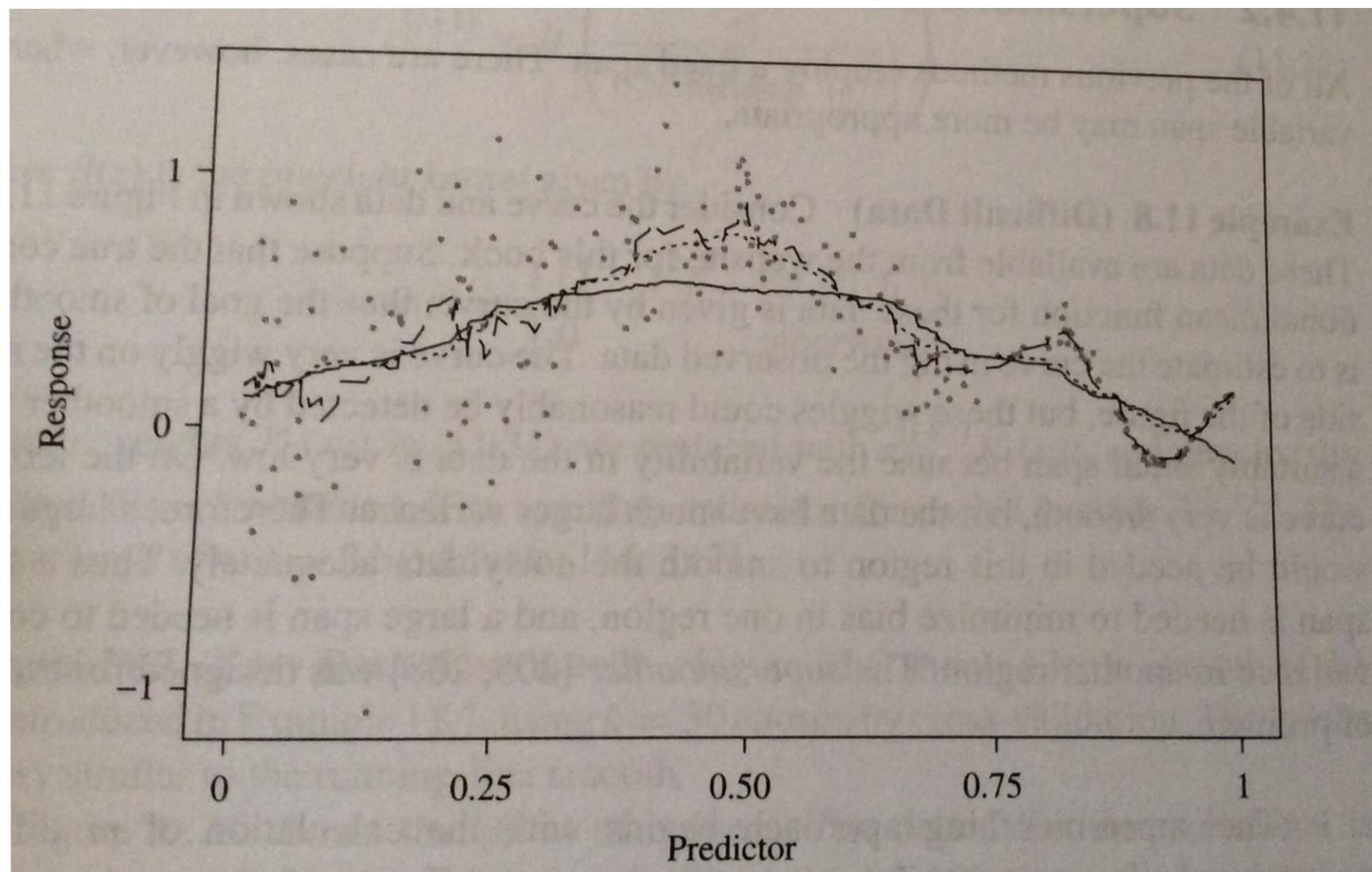
left side
⇒ large span

right side
⇒ small span

need 0
variable
span

Supersmoothing Approach

Step 0: Calculate m different smooths
 $\hat{S}_1(x), \dots, \hat{S}_m(x)$ each with a different
fixed span h_1, \dots, h_m



$$m = 3$$

$$h_1 = .05n$$

$$h_2 = .2n$$

$$h_3 = .5n$$

Supersmoothing Approach

Step 2: Define $P(h_j, x)$ to be a measure of performance of the j th smooth at point x

I ideally we would use $E[g(\gamma - \hat{s}_j^{(i)}) | X = x_i]$

$$\hat{P}(h_j, x_i) = \hat{s}^*(g(\gamma_i - \hat{s}_j^{(i)})(x_i)) \text{ where}$$

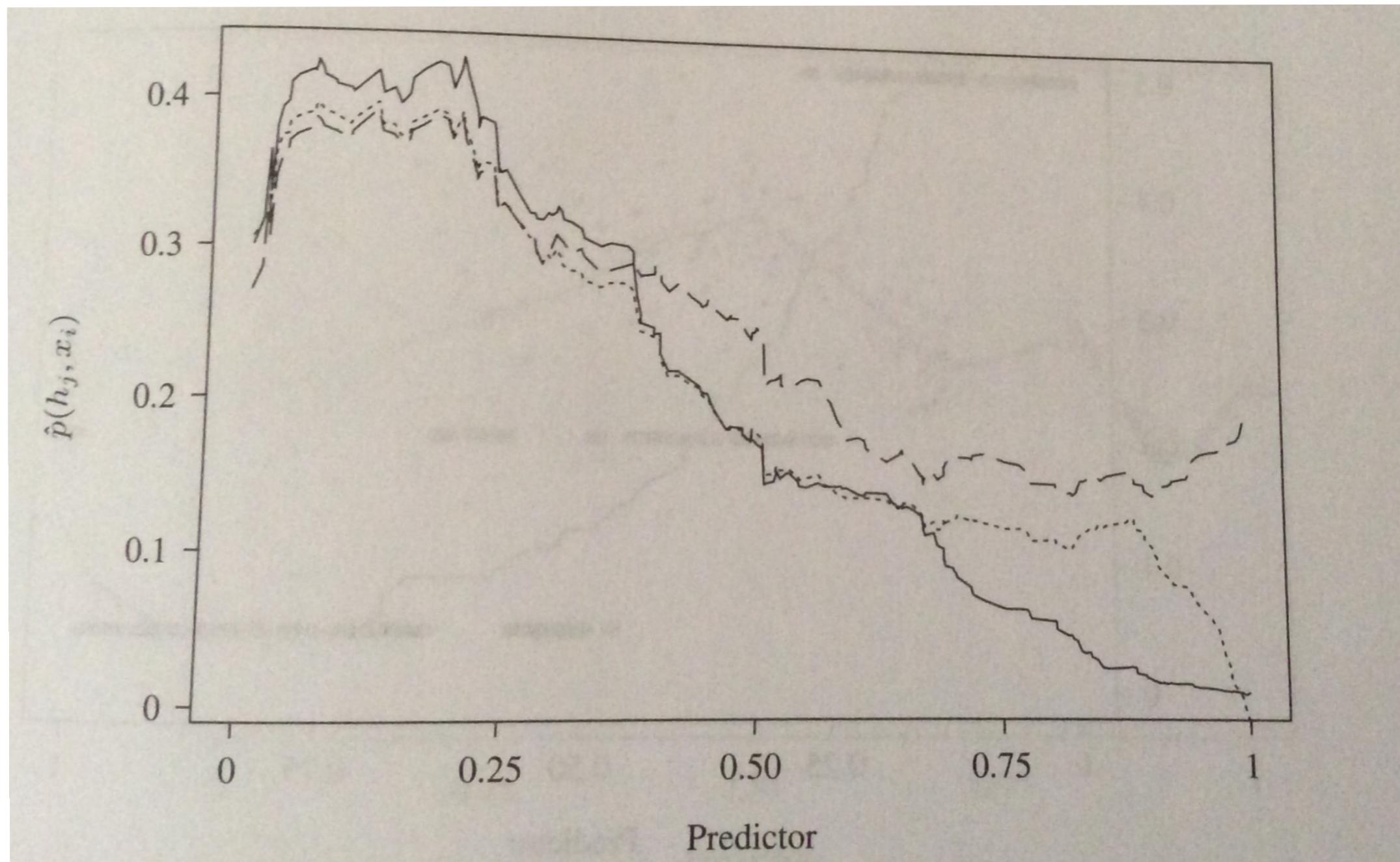
s^* is a fixed span smoother

Step 2 Example

$$\text{Ex } \hat{S}^* = \hat{\Sigma}$$

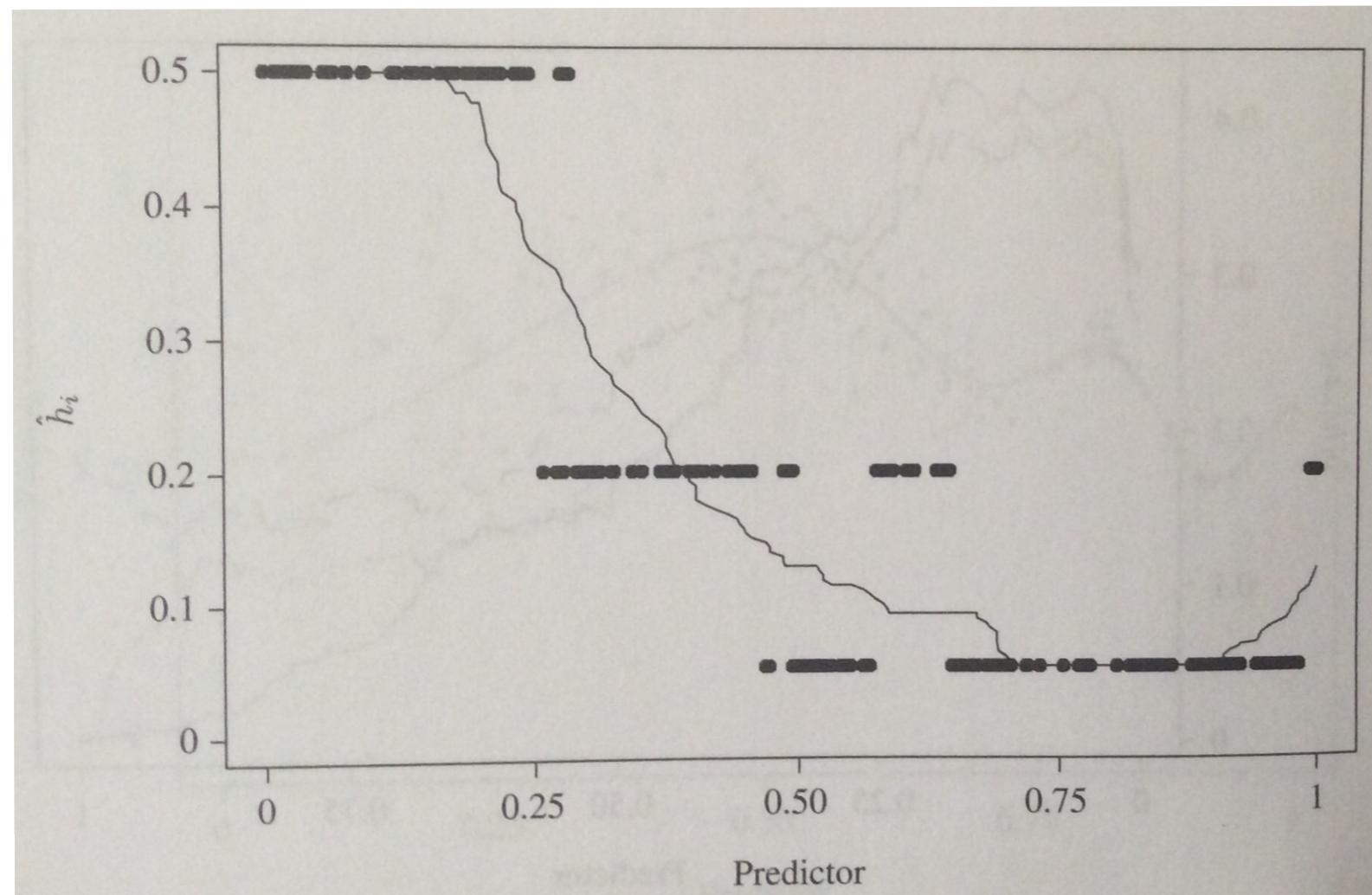
$$g(x) = |x|$$

--- $h_3 = .5n$
.... $h_2 = .2n$
— $h_1 = .05n$



Supersmoothing Approach

Step 3: At each x_i denote the best of the spans as \hat{h}_i — ie the span w/the lowest $\hat{p}(h_j, x_i)$



- pass (x_i, \hat{h}_i) through some smoother \hat{s}^* to obtain optimal span $\hat{h}(x)$ as a function of x

Final Smooth

The final step is to use a linear interpolation between $\hat{s}_{h^-(x_i)}(x_i)$ and $\hat{s}_{h^+(x_i)}(x_i)$ where among the m fixed spans

$h^-(x_i)$ is the largest span $\hat{s}_{h^-(x_i)}$

$h^+(x_i)$ is the smallest span $\hat{s}_{h^+(x_i)}$

Thus

$$\hat{s}(x_i) = \frac{h(x_i) - h^-(x_i)}{h^+(x_i) - h^-(x_i)} \hat{s}_{h^+(x_i)}(x_i) + \frac{h^+(x_i) - h(x_i)}{h^+(x_i) - h^-(x_i)} \hat{s}_{h^-(x_i)}(x_i)$$

Supersmoothing Final Example

