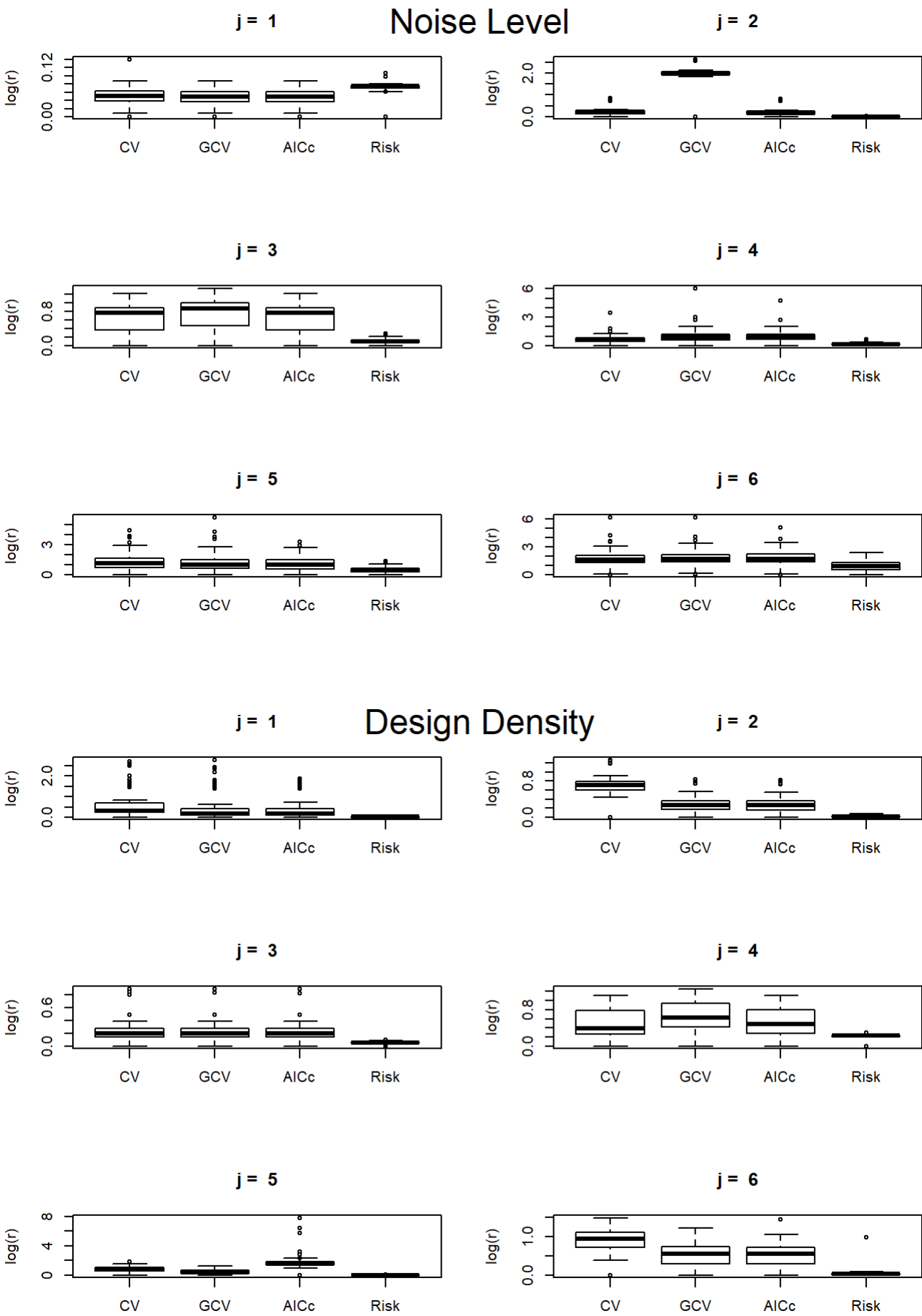


# HW5\_Maheshwari

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## Boxplots







Solution:-

a), b) Penalized Regression splines were fitted using the experimental setup attached in the paper. For all four different setups, namely, Noise level, Design density, spatial variation and Variance function, 6 different conditions were considered.  $\lambda$  for each of these setups was optimized using crossvalidation, ~~and~~ AICc and generalized crossvalidation approaches. Finally, a boxplot of  $\log_e r$  was created for each distribution, where,

$$r = \frac{\|f - \hat{f}_\lambda\|^2}{\min_\lambda \|f - \hat{f}_\lambda\|^2}$$

~~c)~~

c) To prove:  $E\|y - \hat{f}_\lambda\|^2 = \|(I - H_\lambda)f\|^2 + \sigma^2 \{ \text{tr}(H_\lambda H_\lambda^T) - 2\text{tr}(H_\lambda) + n \}$

$$\begin{aligned} \text{L.H.S: } E\|y - \hat{f}_\lambda\|^2 &= E\|y - H_\lambda y\|^2 = E\|(f + e) - H_\lambda(f + e)\|^2 \\ &= E\|(I - H_\lambda)f + (I - H_\lambda)e\|^2 \\ &= \|(I - H_\lambda)f\|^2 + E(\|(I - H_\lambda)e\|^2) + 2[E(I - H_\lambda)f \cdot (I - H_\lambda)e] \end{aligned}$$

$$\begin{aligned} &= \|(I - H_\lambda)f\|^2 + E(e^T (I - H_\lambda)^T (I - H_\lambda) e) \\ &= \|(I - H_\lambda)f\|^2 + E(\text{tr}(e^T (I - H_\lambda)^T (I - H_\lambda) e)) \\ &= \|(I - H_\lambda)f\|^2 + E(\text{tr}((I - H_\lambda)^T (I - H_\lambda) e e^T)) \\ &= \|(I - H_\lambda)f\|^2 + \sigma^2 \cdot E(\text{tr}((I - H_\lambda)^T (I - H_\lambda))) \\ &= \|(I - H_\lambda)f\|^2 + \sigma^2 \{ n - 2\text{tr}(H_\lambda) + \text{tr}(H_\lambda H_\lambda^T) \} \end{aligned}$$



To estimate the risk estimator

$$\text{risk}(\lambda) = E \|f - f_\lambda\|^2$$

$$= E \|f - H_\lambda y\|^2$$

$$= E \|f - H_\lambda f - H_\lambda e\|^2$$

$$= E \|(I - H_\lambda)f - H_\lambda e\|^2$$

$$= \|(I - H_\lambda)f\|^2 + E \|H_\lambda e\|^2 - 2 E \|(I - H_\lambda)f \cdot H_\lambda e\|$$

$\downarrow 0$

$$= \|(I - H_\lambda)f\|^2 + E(\text{tr}(e^T H_\lambda^T H_\lambda e))$$

$$= \|(I - H_\lambda)f\|^2 + \sigma^2 \text{tr}(H_\lambda H_\lambda^T)$$

$$\text{Thus, risk}(\lambda) = E \|y - \hat{f}_\lambda\|^2 - \sigma^2 \{n - 2\text{tr}(H_\lambda) + \text{tr}(H_\lambda H_\lambda^T)\} \\ + \sigma^2 \text{tr}(H_\lambda H_\lambda^T)$$

$$= E \|y - \hat{f}_\lambda\|^2 - \sigma^2 \{n - 2\text{tr}(H_\lambda)\}$$

Also,  $\hat{\sigma} = \sqrt{\frac{\sum (y - \hat{f})^2}{n-1}}$  is considered as a reasonable estimate for  $\sigma$ .

This approach has been implemented in R, the code has been attached and the boxplots are prepared for comparing the  $\log_e$  values.