Sparse Envelope Model (Spenv)

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Fission yeast data

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Cell Cycle of Fission Yeast





- Data: Microarray time-course data [Rustici et al., 2004] on cell-cycle control in the fission yeast
 - Y: expression levels of 396 genes
 - X: 10 equally spaced time points of the cell cycle
 - n: 178 sample hybridizations

 Purpose: study how gene expression levels change in a cell cycle

Envelope Method

■ Under the Multivariate linear regression model

$$\mathsf{Y} = eta \mathsf{X} + \varepsilon, \quad \varepsilon \sim (\mathbf{0}, \mathbf{\Sigma})$$

where $\mathsf{Y} \in \mathbb{R}^r$, $\mathsf{X} \in \mathbb{R}^p$, and $\boldsymbol{\beta} \in \mathbb{R}^{r imes p}$

• Given an orthogonal matrix $(\Gamma, \Gamma_0) \in \mathbb{R}^{r \times r}$, where $\Gamma \in \mathbb{R}^{r \times u}$, then

$$\mathbf{Y} = (\mathsf{P}_{\Gamma} + \mathsf{Q}_{\Gamma})\mathbf{Y} = \mathsf{P}_{\Gamma}\mathbf{Y} + \mathsf{Q}_{\Gamma}\mathbf{Y}$$

where $P_{\Gamma} = \Gamma \Gamma^{T}$ and $Q_{\Gamma} = I_{r} - P_{\Gamma} = \Gamma_{0} \Gamma_{0}^{T}$ are the projection matrices.

• Goal of envelope: $P_{\Gamma}Y$ contains all the information in Y that can help us estimate β .

Envelope Method

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- Model assumptions:
- (1) $\mathbf{Q}_{\Gamma}\mathbf{Y}|\mathbf{X} \sim \mathbf{Q}_{\Gamma}\mathbf{Y}$ \Leftrightarrow $\boldsymbol{\beta} = \boldsymbol{\Gamma}\boldsymbol{\eta}$ (2) $\operatorname{Cor}(\mathbf{P}_{\Gamma}\mathbf{Y}, \mathbf{Q}_{\Gamma}\mathbf{Y} \mid \mathbf{X}) = 0$ \Leftrightarrow $\boldsymbol{\Sigma} = \boldsymbol{\Gamma}\boldsymbol{\Omega}\boldsymbol{\Gamma}^{T} + \boldsymbol{\Gamma}_{0}\boldsymbol{\Omega}_{0}\boldsymbol{\Gamma}_{0}^{T}$

• Coordinate form of the envelope model:

$$\mathbf{Y} = \frac{\mathbf{\Gamma}\boldsymbol{\eta}}{\mathbf{X}} \mathbf{X} + \boldsymbol{\varepsilon}, \quad \mathbf{\Sigma} = \mathbf{\Gamma}\boldsymbol{\Omega}\mathbf{\Gamma}^{\mathcal{T}} + \mathbf{\Gamma}_{0}\boldsymbol{\Omega}_{0}\mathbf{\Gamma}_{0}^{\mathcal{T}}$$

- Response feature selection
 - Y_i is called **static** if $\beta_i = 0$, otherwise called **dynamic**
 - Y_i is called **inactive** if $\Gamma_i = 0$, otherwise called **active**
- Relationship:
 - Because β = Γη, inactive responses are static. However, an active response may also be static
 - **Proposition**: If Y_i is active and static, then Y_i must be connected with dynamic response (in the covariance graph)
 - This indicates that if an active response is static, it still offers information in estimating the non-zero regression coefficients.

Sparse Envelope Method

- So we try to identify inactive responses.
- Sparse Envelope Model [Su et al., 2016]

$$\mathsf{Y} = \frac{\mathbf{\Gamma}\boldsymbol{\eta}}{\mathbf{\Gamma}}\mathsf{X} + \varepsilon, \quad \mathbf{\Sigma} = \frac{\mathbf{\Gamma}\boldsymbol{\Omega}\mathbf{\Gamma}^{\mathsf{T}} + \mathbf{\Gamma}_{0}\boldsymbol{\Omega}_{0}\mathbf{\Gamma}_{0}^{\mathsf{T}}}{\mathbf{\Gamma}}, \quad \mathbf{\Gamma} = \begin{pmatrix} \mathbf{\Gamma}_{\mathcal{A}} \\ \mathbf{0} \end{pmatrix}$$

- Asymptotic Properties:
 - Selection consistency: probability of selection inactive responses goes to 1
 - Asymptotic efficiency: estimator has the same asymptotic covariance matrix as the oracle estimator

Fission yeast data



Spenv identified 25 inactive responses (genes whose intensities do not change in a cell cycle).

Some of them are also confirmed by other researchers: gene cdc20 was also identified by [Gilks et al., 2005] to have "very little cell-cycle activity"

References

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