

# Rosen's paper review

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## 1 Model Specifications

$$\mathbf{y}_i(t_{ij}) = \mathbf{X}_{ij}\boldsymbol{\mu}(t_{ij}) + \mathbf{Z}_{ij}\mathbf{g}_i(t_{ij}) + \boldsymbol{\delta}_i(t_{ij}) \quad (1)$$

where

$$\mathbf{y}_i(t_{ij}) : p \times 1 \quad i = 1, \dots, n, j = 1, \dots, m_i \quad (2)$$

$$\boldsymbol{\mu}(t) = (\boldsymbol{\mu}'_1(t), \dots, \boldsymbol{\mu}'_p(t))' \quad (3)$$

$$\boldsymbol{\mu}'_k(t) = (\mu_{k1}(t), \dots, \mu_{kr}(t))' : \quad r \times 1 \quad (4)$$

$$\mathbf{g}_i(t) = (\mathbf{g}'_{i1}(t), \dots, \mathbf{g}'_{ip}(t))' \quad (5)$$

$$\mathbf{g}_{ik}(t) = (g_{ij1}(t), \dots, g_{iks}(t))' : \quad s \times 1 \quad (6)$$

$$\mathbf{x}_{ij} : r \times 1 \quad (7)$$

$$\mathbf{z}_{ij} : s \times 1 \quad (8)$$

$$\mathbf{X}_{ij} = \mathbf{I}_p \otimes \mathbf{x}'_{ij} \quad (9)$$

$$\mathbf{Z}_{ij} = \mathbf{I}_p \otimes \mathbf{z}'_{ij} \quad (10)$$

$$\boldsymbol{\delta}_i(t_{ij}) \sim \text{Ornstein-Uhlenbeck process} \quad (11)$$

## 2 Ornstein-Uhlenbeck process

An Ornstein-Uhlenbeck process is the second-order stationary process  $\{X_t\}$ , that satisfies the following differential equation:

$$dX(t) = -aX(t) dt + \sigma dB(t), \quad t \geq 0 \quad (12)$$

where  $\{B(t)\}$  is standard Brownian motion, and  $a$  and  $\sigma > 0$  are parameters and  $X_0$  is a random variable that is independent of  $\{B(t)\}$ .

### 2.1 Multivariate Ornstein-Uhlenbeck process

The univariate version of the OU process naturally evolves to a multivariate form which reads as follows:

$$d\mathbf{X}(t) = -\mathbf{A}\mathbf{X}(t) dt + \mathbf{B} d\mathbf{W}(t), \quad (13)$$

where  $\mathbf{A}, \mathbf{B}$  are constant matrices. The solution for this SDE is

$$\mathbf{X}(t) = \exp(-\mathbf{A}t) \mathbf{X}_0 + \int_0^t \exp\{-\mathbf{A}(t-t')\} \mathbf{B} d\mathbf{W}(t'). \quad (14)$$

The properties of such an OU process are as follows:

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- If  $\mathbf{A}$  has only eigenvalues with positive real part, a stationary solution exists of the form

$$\mathbf{X}_s(t) = \int_{-\infty}^t \exp\{-\mathbf{A}(t-t')\} \mathbf{B} d\mathbf{W}(t'). \quad (15)$$

The expected value  $E[\mathbf{X}_s(t)] = 0$  and the covariance matrix

$$\mathbf{\Sigma} = \text{Cov}(\mathbf{X}_s(t), \mathbf{X}'_s(s)) = \int_{-\infty}^{\min(t,s)} \exp\{-\mathbf{A}(t-t')\} \mathbf{B} \mathbf{B}' \exp\{-\mathbf{A}'(s-t')\} dt'. \quad (16)$$

- The stationary covariance matrix satisfies the following equation:

$$\mathbf{A}\mathbf{\Sigma} + \mathbf{\Sigma}\mathbf{A}' = \mathbf{B}\mathbf{B}'. \quad (17)$$

The solution to this equation is given by

$$\mathbf{\Sigma} = \frac{|\mathbf{A}| \mathbf{B} \mathbf{B}' + [\mathbf{A} - \text{Tr}(\mathbf{A}) \mathbf{I}] \mathbf{B} \mathbf{B}' [\mathbf{A} - \text{Tr}(\mathbf{A}) \mathbf{I}]}{2(\text{Tr}(\mathbf{A})) |\mathbf{A}|} \quad (18)$$

- (*Time Correlation Matrix in the Stationary State*) The following relations hold.

$$\text{Cov}(\mathbf{X}_s(t), \mathbf{X}'_s(s)) = \exp\{-\mathbf{A}(t-s)\} \mathbf{\Sigma}, \quad t > s \quad (19)$$

$$= \mathbf{\Sigma} \exp\{-\mathbf{A}'(s-t)\}, \quad t < s \quad (20)$$

- If all its transition densities depend only on the time differences, then the Markov process is *homogeneous*. OU process is homogeneous. Therefore, the transition probability of an OU process is given by

$$P(\mathbf{X}_s(t) | \mathbf{X}_s(s), t-s) = |2\pi\mathbf{\Omega}|^{-1/2} \exp\left\{-\frac{1}{2}\boldsymbol{\gamma}'\mathbf{\Omega}^{-1}\boldsymbol{\gamma}\right\} \quad (21)$$

where

$$\boldsymbol{\gamma} = \mathbf{X}_s(t) - \exp(-\mathbf{A}(t-s)) \mathbf{X}_s(s) \quad (22)$$

$$\mathbf{\Omega} = \mathbf{\Sigma} - \exp\{-\mathbf{A}(t-s)\} \mathbf{\Sigma} \exp\{-\mathbf{A}'(t-s)\}. \quad (23)$$

### 3 Cubic Spline