# Non-linear Spatio-temporal modelling using the Integro-Difference Equation

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#### **Abstract**

### **Index Terms**

Dynamic spatio-temporal modelling, Integro-Difference Equation (IDE), ...

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# Non-linear Spatio-temporal modelling using the Integro-Difference Equation

### I. Introduction

Non-linear spatio-temporal behaviour...

#### II. NEURAL FIELD MODEL

### III. STATE SPACE REPRESENTATION OF THE SPATIO-TEMPORAL NON-LINEAR IDE

We define the pth order non-linear, stochastic, spatio-temporal IDE as

$$\rho_t(\mathbf{r}) = \int_{\Omega} w(\mathbf{r}, \mathbf{r}') f(v_{t-p}(\mathbf{r}')) d\mathbf{r}' + \mathbf{e}_{t-p}(\mathbf{r})$$
(1)

where  $\rho_t(\mathbf{r})$  is a lag polynomial defined as

$$\rho_t(\mathbf{r}) = \left(1 + \sum_{i=1}^p \varphi_i L_i\right) v_t(\mathbf{r}) \tag{2}$$

where  $v_t(\mathbf{r})$  denotes the spatial field at time t and at spatial location  $\mathbf{r}$ , f(.) is some nonlinear map, and w(.) is the spatial kernel. The disturbance  $\mathbf{e}_{t-p}(\mathbf{r})$  is a zero-mean normally distributed noise process, spatially coloured but temporally independent, with covariance

$$cov\left(e_{t}\left(\mathbf{r}\right), e_{t+\tau}\left(\mathbf{r}'\right)\right) = \gamma\left(\mathbf{r} - \mathbf{r}'\right)\delta(t - \tau)$$
(3)

where  $\delta(.)$  is the Dirac-delta function. The spatial field is observed by

$$\mathbf{y}_{t} = \int_{\Omega} m \left( \mathbf{r}_{n_{y}} - \mathbf{r}' \right) v_{t} \left( \mathbf{r}' \right) d\mathbf{r}' + \boldsymbol{\varepsilon}_{t}, \tag{4}$$

where the observation vector  $\mathbf{y}_t$ , compiled at  $n_y$  spatial locations and at time t, is corrupted by an i.i.d normally distributed zero-mean white noise  $\boldsymbol{\varepsilon}_t \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Sigma}_{\varepsilon}\right)$  with  $\boldsymbol{\Sigma}_{\varepsilon} = \sigma_{\varepsilon}^2 \mathbf{I}_{n_y}$ . It is assumed here that  $\boldsymbol{\Sigma}_{\varepsilon}$  and  $\gamma\left(\mathbf{r} - \mathbf{r}'\right)$  are known. The output kernel  $m(\mathbf{r} - \mathbf{r}')$  governs the sensor pick-up geometry and is defined by

$$m(\mathbf{r} - \mathbf{r}') = \exp\left(-\frac{(\mathbf{r} - \mathbf{r}')^{\top}(\mathbf{r} - \mathbf{r}')}{\sigma_m^2}\right),$$
 (5)

where  $\sigma_m$  sets the sensor width. The superscript  $\top$  denotes the transpose operator.

Here we consider the case where p = 1, therefore we have

$$v_t(\mathbf{r}) + \varphi_1 v_{t-1}(\mathbf{r}) = \int_{\Omega} w(\mathbf{r}, \mathbf{r}') f(v_{t-1}(\mathbf{r}')) d\mathbf{r}' + \mathbf{e}_{t-1}(\mathbf{r})$$
(6)

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## IV. ESTIMATION OF THE NON-LINEAR SPATIO-TEMPORAL IDE MODEL

- A. Unscented Kalman Filter
- B. Unscented RTS Smoother
- C. Parameter Estimation
- D. Computational Complexity

## V. EXAMPLE

## VI. CONCLUSION

## APPENDIX

Lemma 1:

Proof:

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