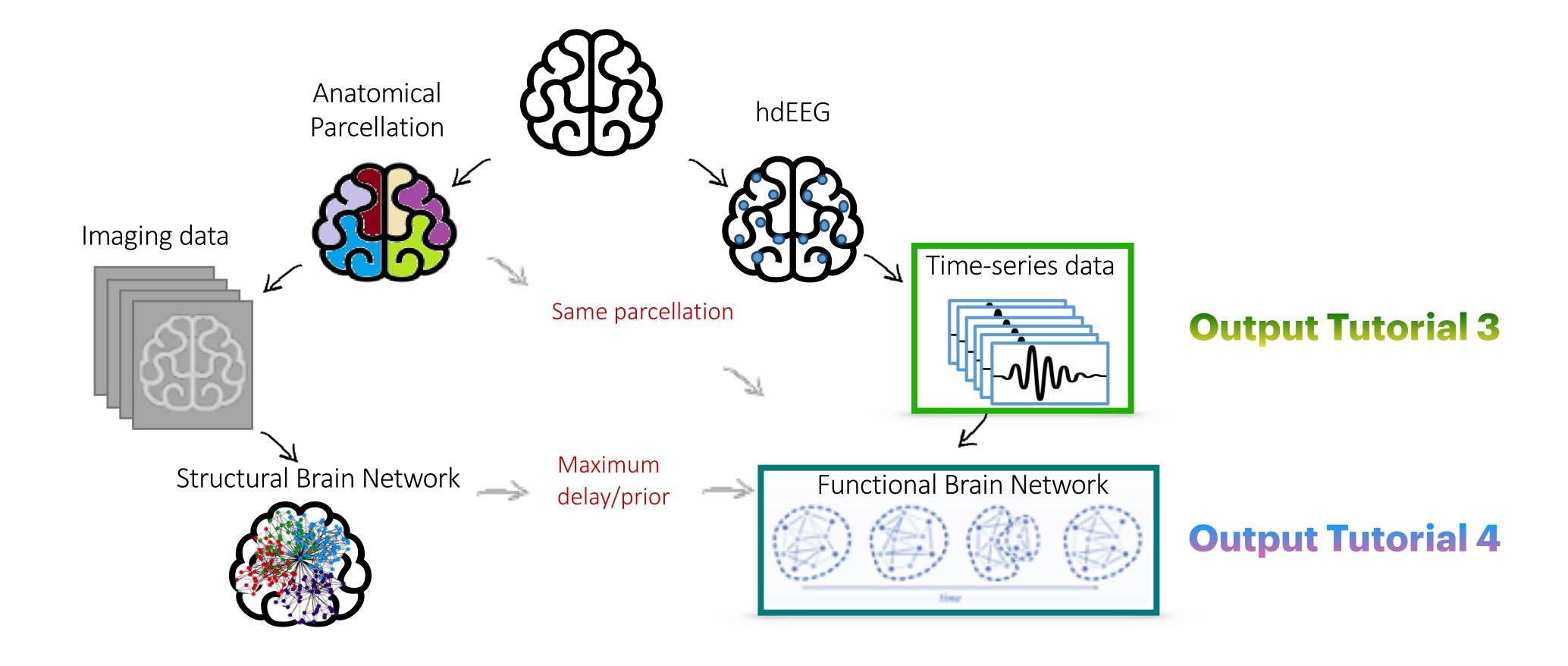
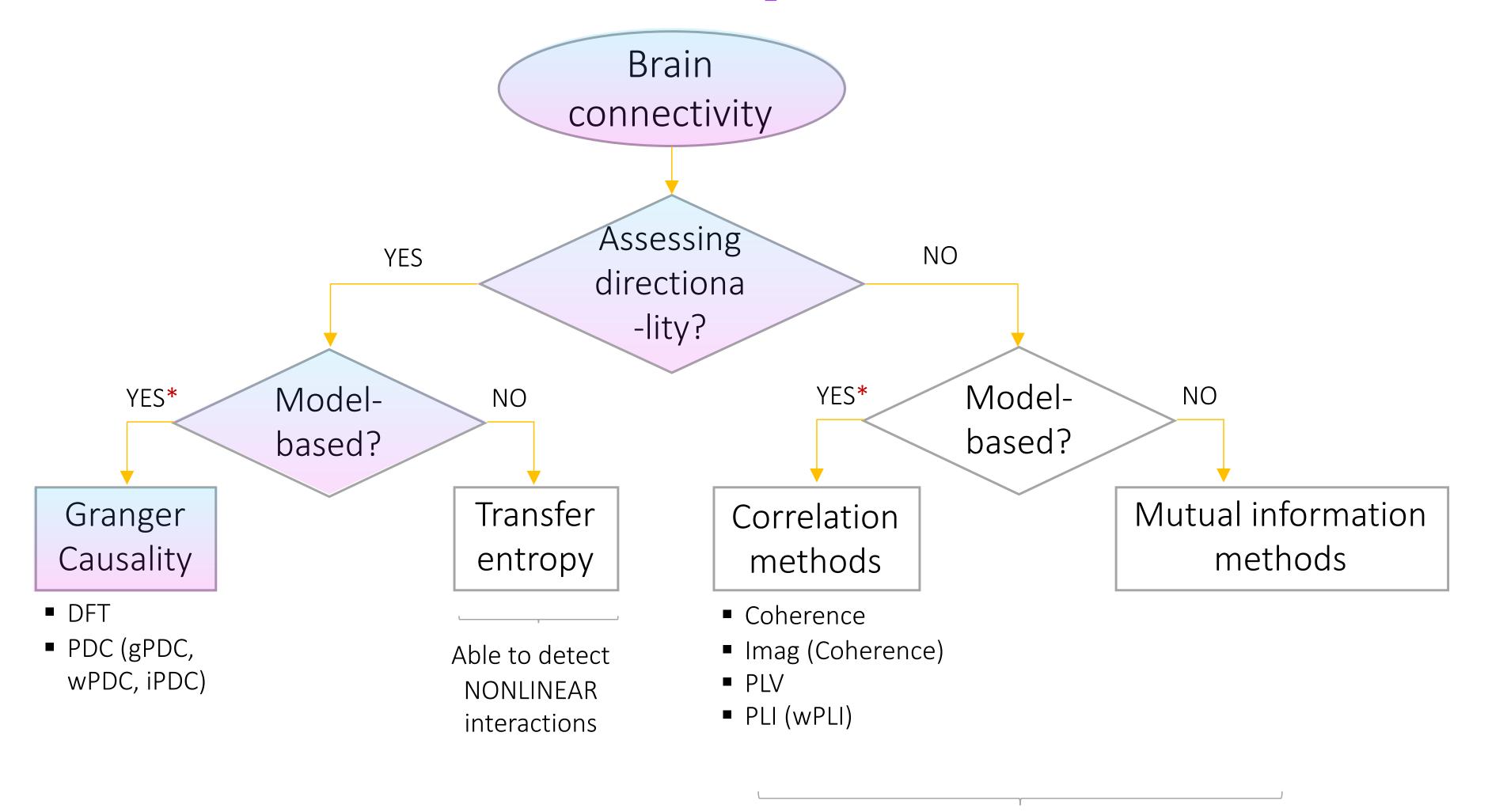
Dynamic Brain Connectivity Tutorial 4

Extraction of brain network



Brain connectivity methods

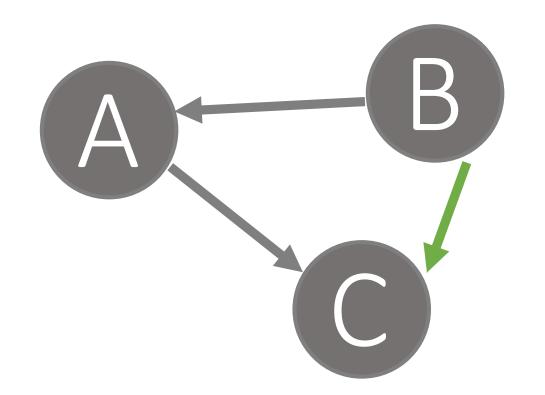


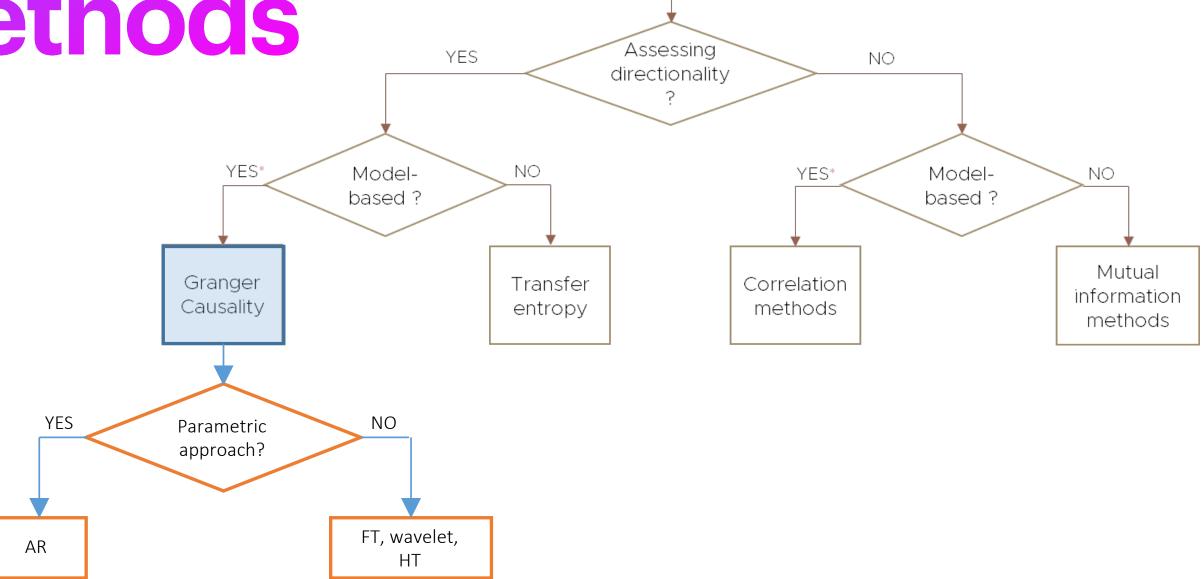
^{*} Assumption of linearity

Ignore temporal structure of the data

Effective connectivity methods

Granger causality





Functional

connectivity

 Direct Transfer function (DTF) → Impossible to distinguish direct or not-direct influences

Comparable results

2. Partial Directed Coherence (PDC) → Estimation of only direct influences

Sameshima, K., & Baccala, LA. (2016). CRC press

Assumption

The cortical sources computed from the EEG data generate a collection of realisations of a multivariate stochastic process which can be combined in a multivariate, multi-trial time $Y_k = \begin{bmatrix} y_{1,k}^{(1)} & \cdots & y_{d,k}^{(1)} \\ \vdots & \ddots & \vdots \\ y_{1,k}^{(N)} & \cdots & y_{d,k}^{(N)} \end{bmatrix} \quad k = t_1, \dots, t_T$ series.

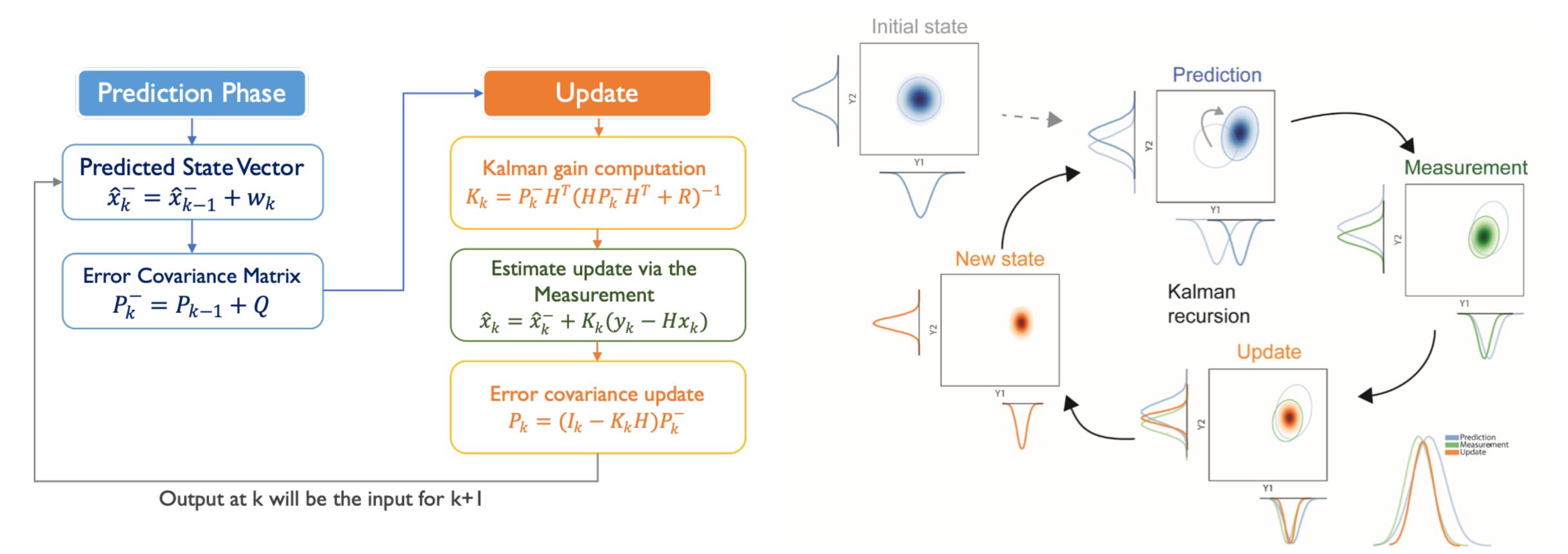
$$Y_{k} = \begin{bmatrix} y_{1,k}^{(1)} & \cdots & y_{d,k}^{(1)} \\ \vdots & \ddots & \vdots \\ y_{1,k}^{(N)} & \cdots & y_{d,k}^{(N)} \end{bmatrix} \quad k = t_{1}, \dots, t_{T}$$

The process is defined by an integer order parameter \mathbf{p} , a stationary d-dimensional white $Y_k = \sum_{n=1}^{P} X_{n,k} Y_{k-n} + \varepsilon_k$ The process is defined by an integer order noise process and p parameter matrices at each time step, the so-called AR matrices. Nonzero entries in the AR matrices at (i,j) describe connections from source j to source i.

$$Y_k = \sum_{n=1}^p X_{n,k} Y_{k-n} + \varepsilon_k$$

Rubega et al, 2018; Milde et al, 2010

Real time Kalman filtering



Milde et al, 2010

Image credit: David Pascucci | EPFL

Parameters to set

$$Y_k = \sum_{n=1}^{p} X_{n,k} Y_{k-n} + \varepsilon_k$$

$$K_k = P_k^- H^T (H P_k^- H^T + R)^{-1}$$

$$\hat{R}_t = \hat{R}_{t-1} + c(\Sigma_r - \hat{R}_{t-1})$$

 Σ_r : measurement innovation covariance matrix

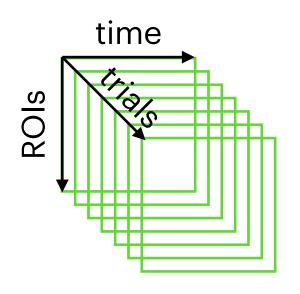
p-order of the multi-variate autoregressive model

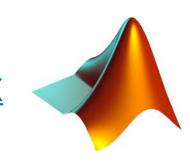
Adaptation constant c of the Kalman gain

p-order of the multi-variate autoregressive model

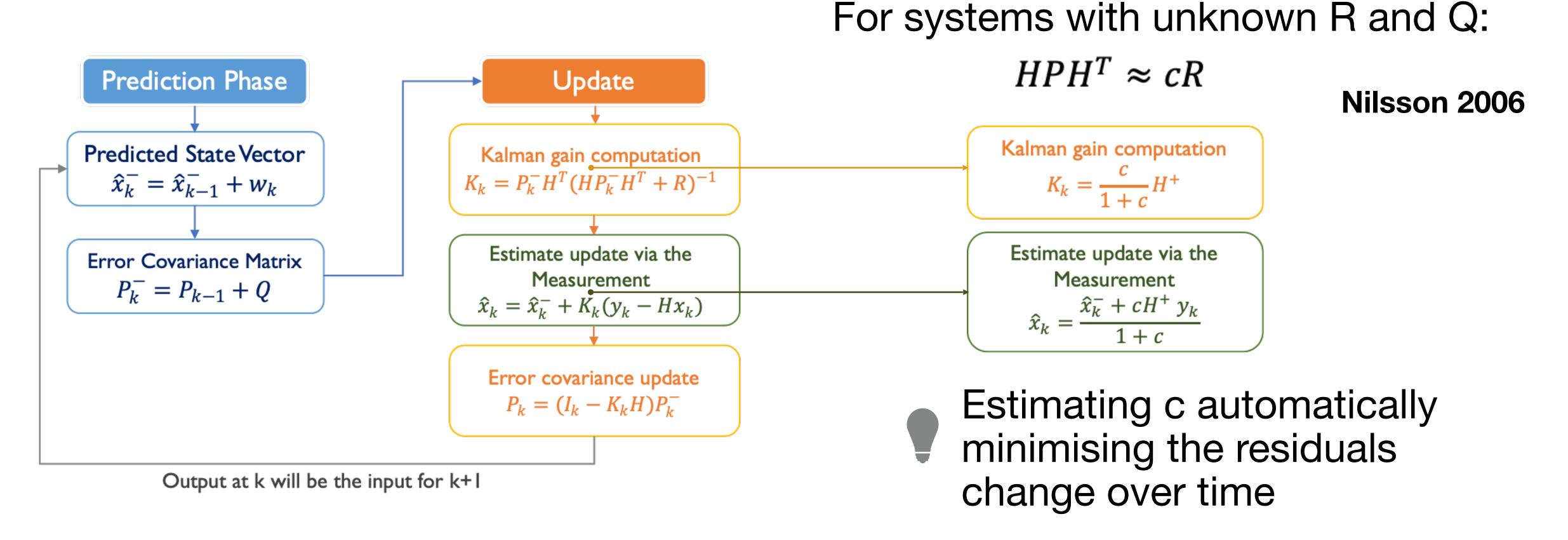
Adaptation constant c of the Kalman gain

```
p_value=6;
C=0.02;
temp=VEP_faces(1:30,:,373:501);
t_new=t(373:501);
gKF=dynet_SSM_KF(temp,p_value,C);
```

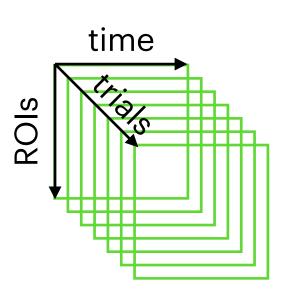




Self-tuning Optimized Kalman (STOK)

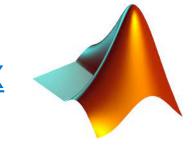


Pascucci, Rubega and Plomp, 2020

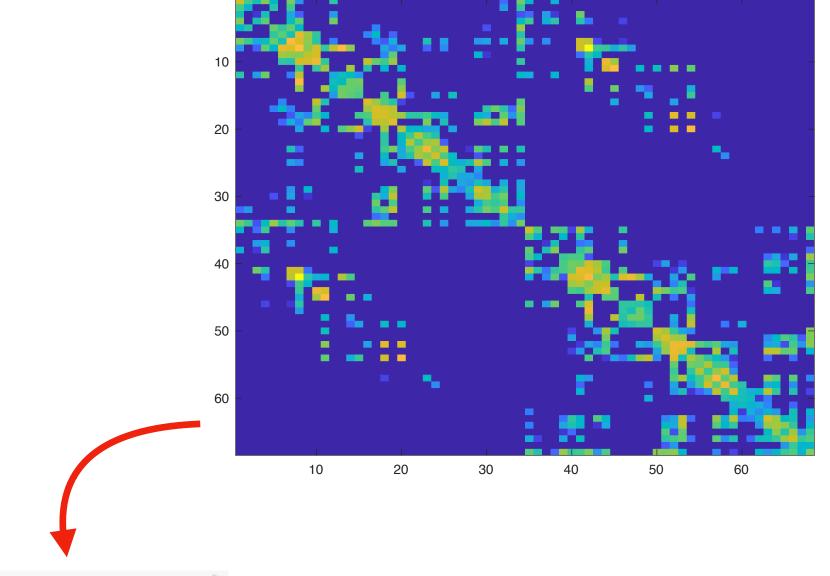


p-order of the multi-variate autoregressive model

STOK=dynet_SSM_STOK(temp,p_value);



Structural prior



siSTOK = dynet_SSM_siSTOK(sitemp,p_order,SC_prior,.99)

