Kalman filter tuning and consistency – Innovation

Sensor fusion & nonlinear filtering

Lars Hammarstrand

INNOVATION CONSISTENCY

Innovation consistency

The innovation $\mathbf{v}_k = \mathbf{y}_k - \mathbf{H}_k \hat{\mathbf{x}}_{k|k-1}$ should satisfy

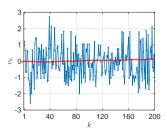
$$p(\mathbf{v}_k | \mathbf{y}_{1:k-1}) = \mathcal{N}(\mathbf{v}_k; \mathbf{0}, \mathbf{S}_k)$$

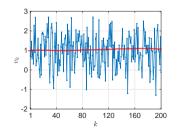
$$\operatorname{Cov}(\mathbf{v}_k, \mathbf{v}_{k-l}) = egin{cases} \operatorname{Cov}\{\mathbf{v}_k\} & \text{if } l = 0 \\ \mathbf{0} & \text{otherwise.} \end{cases}$$

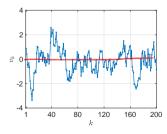
 Note: it is enough to have a filter and a measurement sequence to compute v₁, v₂,...

TEST OF INNOVATION PROPERTIES - VISUAL INSPECTION

- There are ways to test the properties of the innovation.
 - → we will look at three methods.
- Visual inspection:
 - Zero mean?
 - uncorrelated?







TEST OF INNOVATION PROPERTIES - CONSISTENCY

Consistency

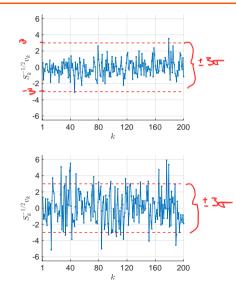
• Ideally $\mathbf{v}_k | \mathbf{y}_{1:k-1} \sim \mathcal{N}(\mathbf{0}, \mathbf{S}_k)$ and then

$$\mathbf{S}_{k}^{-1/2}\mathbf{v}_{k}\sim\mathcal{N}(\mathbf{0},\mathbf{I})\Rightarrow\mathbf{v}_{k}^{T}\mathbf{S}_{k}^{-1}\mathbf{v}_{k}\sim\chi_{n}^{2}$$

Given a sequence v₁, v₂,..., v_K we can compute

$$\xi_{\mathcal{K}} = \sum_{k=1}^{\mathcal{K}} \mathbf{v}_{k}^{\mathsf{T}} \mathbf{S}_{k}^{-1} \mathbf{v}_{k} \sim \mathcal{N}(\mathcal{K} n_{y}, 2\mathcal{K} n_{y})$$

Within 3σ -region?



TEST OF INNOVATION PROPERTIES - CORRELATION

Whiteness

 Estimate the autocorrelation function (autocov. normalised to 1 at lag 0):

$$\rho(l) = \frac{\sum_{k=l+1}^{K} \mathbf{v}_{k}^{T} \mathbf{v}_{k-l}}{\sum_{\tau=l+1}^{K} \mathbf{v}_{\tau}^{T} \mathbf{v}_{\tau}}$$

and check if $\rho(I) \approx 0$ for I > 0.

