

Motivation

- ▶ Partial realization problem:

$$\text{minimize over ?'s rank } \underbrace{\begin{bmatrix} h(1) & h(2) & h(3) & \cdots & h(T) \\ h(2) & h(3) & \ddots & h(T) & ? \\ h(3) & \ddots & \ddots & \ddots & ? \\ \vdots & h(T) & \ddots & \ddots & \vdots \\ h(T) & ? & ? & \cdots & ? \end{bmatrix}}_{\mathcal{H}_T(h)}$$

- ↪ ?'s (missing values) are in the “future”
- ↪ extrapolation of sequence by LTI model

- ▶ Generalization: arbitrary missing values

- ↪ interpolation/extrapolation of a sequence by (multivariable, input/output) LTI model
- ↪ Hankel low-rank matrix completion

$$\begin{aligned} &\text{minimize over } \hat{h} \text{ rank}(\mathcal{H}_L(\hat{h})) \\ &\text{subject to } \hat{h}(\mathcal{I}_{\text{data}}) = h(\mathcal{I}_{\text{data}}) \end{aligned}$$

Solution methods

- ▶ nuclear norm heuristic
- ▶ local optimization methods
 - I. Markovsky and K. Usevich. Structured low-rank approximation with missing values. <http://eprints.soton.ac.uk/340718>*
- ▶ subspace-type method (this poster)

Subspace-type method

- ▶ let $\text{rank}(\mathcal{H}_{r+1}(h)) = r$
- ▶ there is $p \neq 0$, $[p_0 \ p_1 \ \cdots \ p_r] \mathcal{H}_{r+1}(h) = 0$
- ▶ consider the matrix $\mathcal{H}_{r+2}(h)$; we have

$$\underbrace{\begin{bmatrix} p_0 & p_1 & \cdots & p_r & 0 \\ 0 & p_0 & p_1 & \cdots & p_r \end{bmatrix}}_{\tilde{P}} \mathcal{H}_{r+2}(h) = 0$$

- ▶ \tilde{P} is full row rank, so that for any i there is $\tilde{p}^i \neq 0$,
 $\tilde{p}^i \mathcal{H}_{r+2}(h) = 0$ and $\tilde{p}_i^i = 0$

- ▶ suppose that $\mathcal{H}_{r+2}(h)$ has at least $r+1$ columns with single missing value in the i th position and let \tilde{H}^i be the corresponding submatrix of $\mathcal{H}_{r+2}(h)$

- ▶ $\text{left ker}(\tilde{H}^i) = \alpha \tilde{p}^i$, for some $\alpha \neq 0$
 ↪ \tilde{p}^i can be identified (up to scaling factor)

- ▶ suppose also that $\mathcal{H}_{r+2}(h)$ has at least $r+1$ columns with single missing value in the j th position, where $j \neq i$
 ↪ \tilde{p}^j can be identified (up to scaling factor)

- ▶ $\text{GCD}(p^i, p^j) = p$ ↪ identification algorithm with missing values

Applications

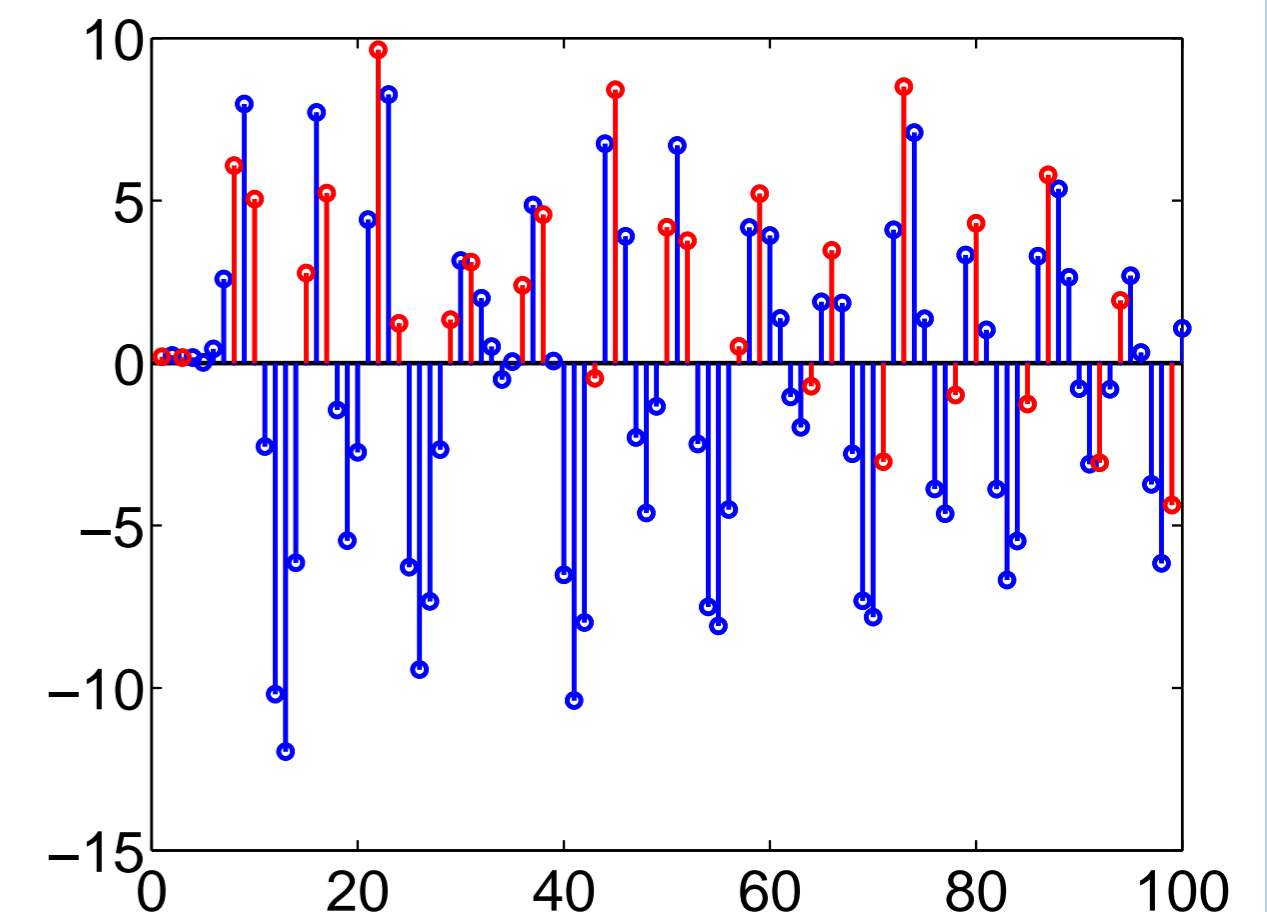
- ▶ system identification with missing data
- ▶ data-driven simulation
- ▶ data-driven tracking

Extensions

- ▶ $\Delta > 1$ missing values per frame
- ▶ multivariable time series
- ▶ input/output systems
- ▶ multiple time series
- ▶ noisy data require approx. kernel and GCD computations

$$\min \left[\begin{array}{l} \text{rank}(\mathcal{H}_L(\hat{h})) \\ \|\hat{h}(\mathcal{I}_{\text{data}}) - h(\mathcal{I}_{\text{data}})\| \end{array} \right] \begin{array}{l} \leftarrow \text{complexity} \\ \leftarrow \text{approx. error} \end{array}$$

Example: exact data, autonomous 6th order system



data points interpolated values

the data is available from

users.ecs.soton.ac.uk/im/d.mat