Analysis of time delay data and clock drift in a network of seismic monitors

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ECMI Modelling Week, 21 July 2018

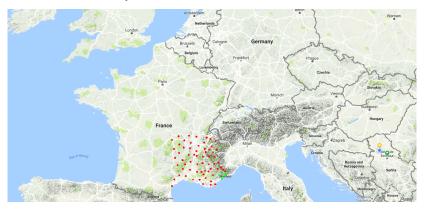


Problem statement

- Seismic monitoring is used to study the behaviour and composition of the underground floor
- ► There is a time drift in the monitors' clocks, causing inaccuracies in analysis
- ► Continuous synchronization to the Global Positioning System for accurate timing is not possible
- A reliable method to correct the time drift of the clocks' is required

The network of seismic monitors

- ▶ The network consists of 73 seismic monitors
- ► For initial analysis, 7 stations are selected

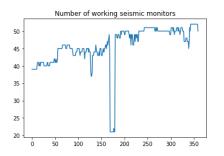


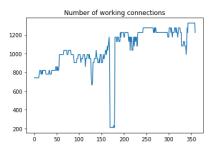
The network of seismic monitors



Numbers of working stations and connections to stations during one year

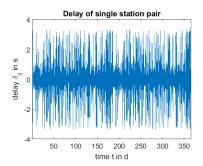
- ► The maximum of simultaneously working stations is 52 out of 73
- The maximum of connections is 1326
- If two monitors are working simultaneously, they are connected

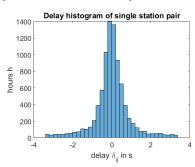




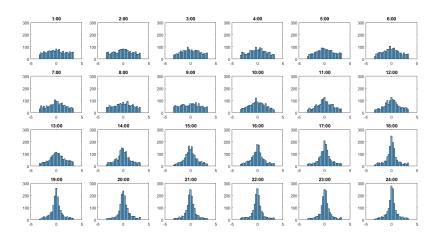
Time-delay signal characteristics

- Data recorded in one second intervals
- ▶ Time-delay cross-correlation in computed in one-hour intervals
- Each connection is computed only once
- Tracing time-delays recorded by one station over a year

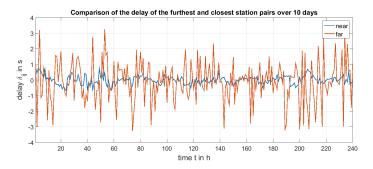




Histograms showing time delays recorded by all active stations over 24 hours



Comparison of time delay of the furthest and the closest station pairs



► Time delays are smaller between stations which are close to each other

Graph approach to model the data

lacktriangle The measurements $\hat{\delta}$ of time delays between stations

$$\hat{\delta} = \delta + \varepsilon$$
 (1)

- lacksquare A weighted graph $\mathcal{G} = (\mathcal{V}, \mathcal{E}, \hat{oldsymbol{\delta}}_t)$
- ▶ Values $\hat{\delta}_{ij}$ are normalized to [0, 1]
- ► A normalized distance measure was selected as graph metrics K_m

$$K_m = f_m(r_i) = \sum_j r_{ij}.$$
 (2)

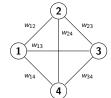
frametitleGraph approach to model the data

lacktriangle The measurements $\hat{\delta}$ of time delays between stations

$$\hat{\delta} = \delta + \varepsilon \tag{3}$$

- lacksquare A weighted graph $\mathcal{G} = (\mathcal{V}, \mathcal{E}, \hat{oldsymbol{\delta}}_t)$
- ▶ Values $\hat{\delta}_{ij}$ are normalized to [0,1]
- ► A normalized distance measure was selected as graph metrics K_m

$$K_m = f_m(r_i) = \sum_i r_{ij}.$$
 (4)



blahblah

Signal denoising

• We specify a cost function $c(\hat{\delta})$ measuring the deviation from the observed weight matrix's metrics $f_m(\hat{\delta})$ to the estimates of the distance metrics K_m

$$c(\hat{\delta}) = \sum_{m} e_m^2(\hat{\delta}) = \sum_{m} (f_m(\hat{\delta}) - K_m)^2$$
 (5)

Error is minimized with gradient descent updates on $\hat{\delta}$:

$$\hat{\delta}^{(t+1)} = \hat{\delta}^t - \mu \sum_{m} e_m(\hat{\delta}^t) \frac{df_m(\hat{\delta}^t)}{d\hat{\delta}^t}, \tag{6}$$

Exctracting clock drift from noise

▶ The noise ε_{ij} between each pair of two stations is modelled with

$$\varepsilon_{ij} = \Delta_i - \Delta_j + e_{ij}, \tag{7}$$

in which Δ_i and Δ_j are the clock drifts of stations i,j e_{ij} accounts for all other noise in the system

Solving individual clock drifts (1/2)

▶ We have overdetermined inverse problem of the form

$$Gm = s$$
 (8)

$$\begin{bmatrix} 1 & -1 & 0 & 0 & \dots & 0 \\ 1 & 0 & -1 & 0 & \dots & 0 \\ 1 & 0 & 0 & -1 & \dots & 0 \\ 1 & 0 & 0 & -1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 0 & 0 & 0 & \dots & -1 \\ 0 & 1 & -1 & 0 & \dots & 0 \\ 0 & 1 & 0 & -1 & \dots & 0 \\ 0 & 1 & 0 & 0 & \ddots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} \Delta_1 \\ \Delta_2 \\ \Delta_3 \\ \Delta_4 \\ \vdots \\ \Delta_k \\ \vdots \\ \Delta_{n-1} \\ \Delta_n \end{bmatrix} = \begin{bmatrix} \Delta_1 - \Delta_2 \\ \Delta_1 - \Delta_3 \\ \Delta_1 - \Delta_4 \\ \vdots \\ \Delta_1 - \Delta_n \\ \Delta_2 - \Delta_3 \\ \Delta_2 - \Delta_4 \\ \vdots \\ \Delta_{2n-1} - \Delta_n \end{bmatrix} = \begin{bmatrix} \Delta_{12} \\ \Delta_{13} \\ \Delta_{14} \\ \vdots \\ \Delta_{2n-1} \\ \Delta_{2n-1} \\ \vdots \\ \Delta_{2n-1} - \Delta_n \end{bmatrix}$$

▶ Matrix G rank is n-1



Solving individual clock drifts 1/2

Ordinary least squares regression

$$\|\mathbf{Gm} - \mathbf{s}\|_2^2. \tag{9}$$

Tikhonov regularization is employed

$$\|\mathbf{Gm} - \mathbf{s}\|_{2}^{2} + \|\mathbf{\Gamma m}\|_{2}^{2},$$
 (10)

for some suitable Tikhonov matrix $\Gamma = \alpha \mathbf{I}$

▶ The explicit solution is hence given by

$$\hat{\mathbf{s}} = (\mathbf{G}^T \mathbf{G} + \mathbf{\Gamma}^T \mathbf{\Gamma})^{-1} \mathbf{G}^T \mathbf{m}. \tag{11}$$



Results

Conclusion