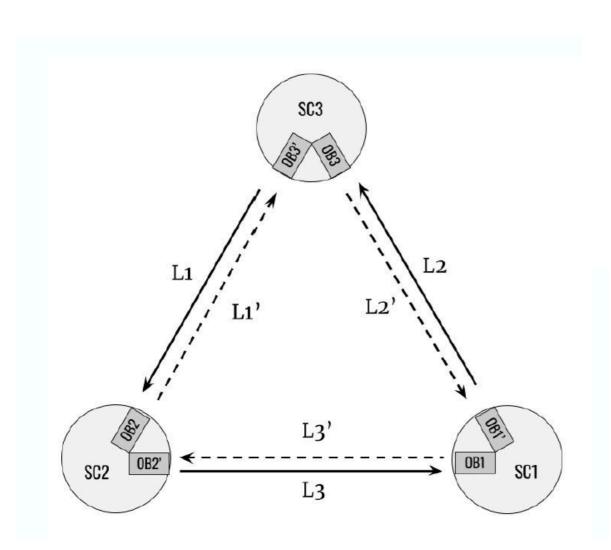
MACHINE LEARNING FOR LISA DATA PREPROCESSING

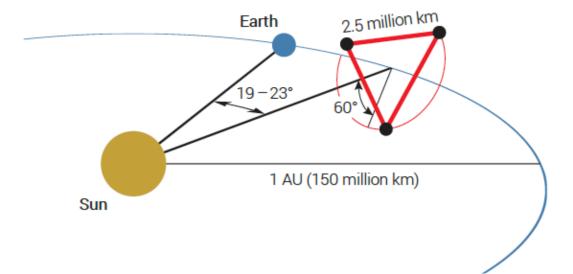
Natalia Korsakova Robyn Munoz

Observatoire de la Côte d'Azur

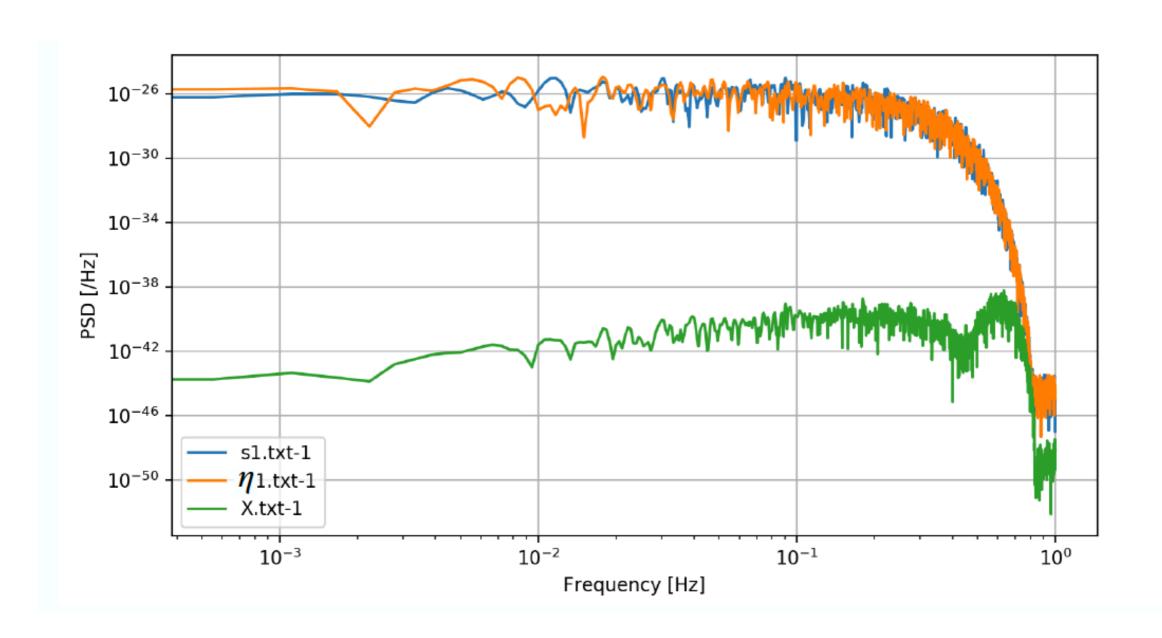


Laser Interferometer Space Antenna

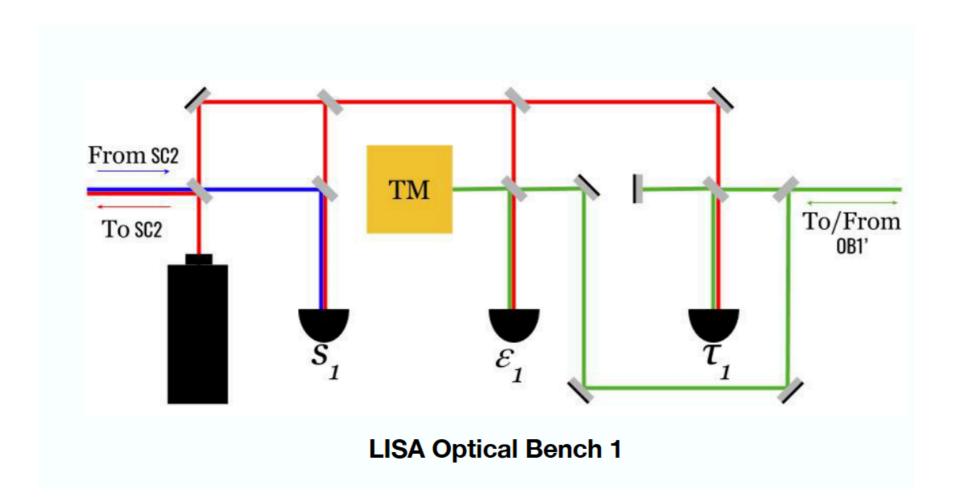




Laser Noise

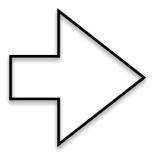


Time Delay Interferometry



Raw Measurements

$$\left\{
\begin{array}{l}
s_{1}, s_{1'}, s_{2}, s_{2'}, s_{3}, s_{3'}, \\
\varepsilon_{1}, \varepsilon_{1'}, \varepsilon_{2}, \varepsilon_{2'}, \varepsilon_{3}, \varepsilon_{3'}, \\
\tau_{1}, \tau_{1'}, \tau_{2}, \tau_{2'}, \tau_{3}, \tau_{3'}
\end{array}
\right\}$$



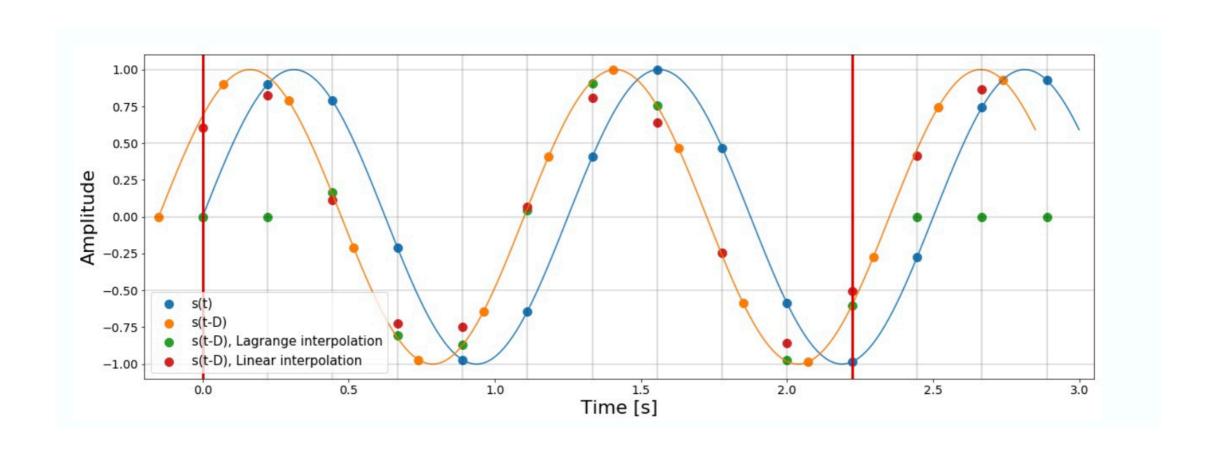
Michelson Combinations

$$\{X, Y, Z\}$$

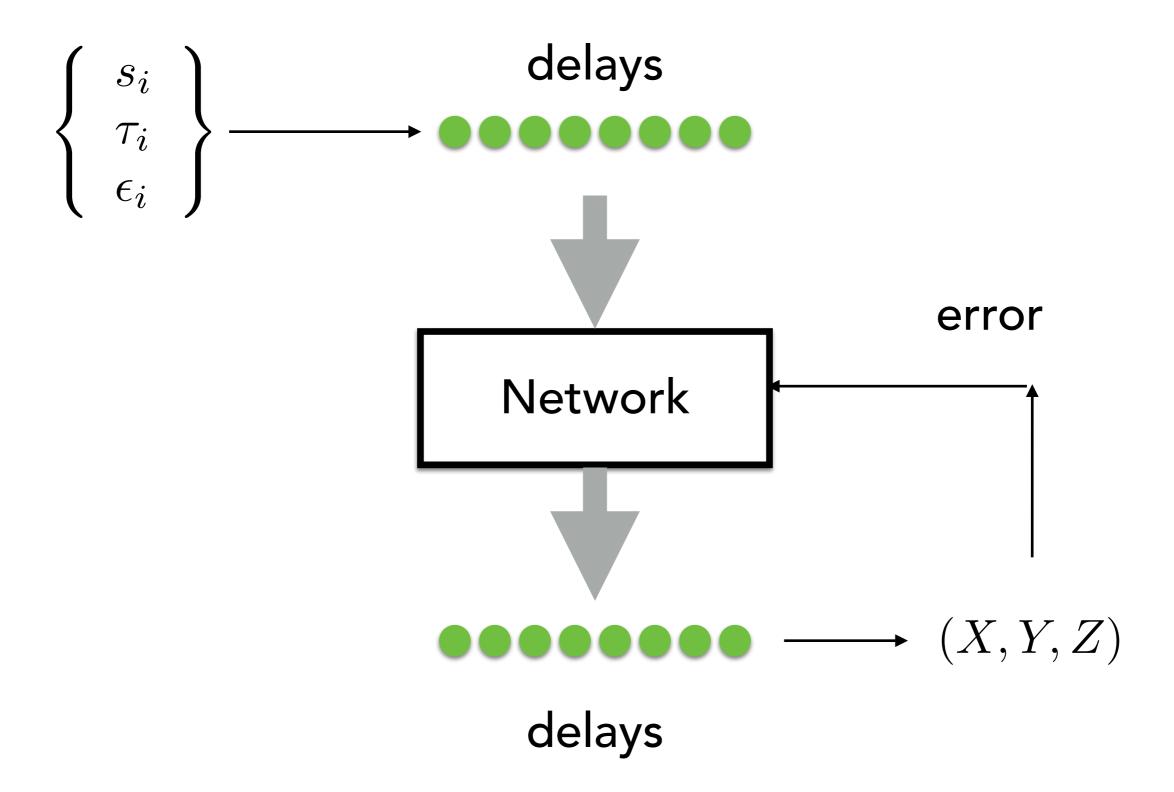
Lagrange Interpolation

Apply time delays

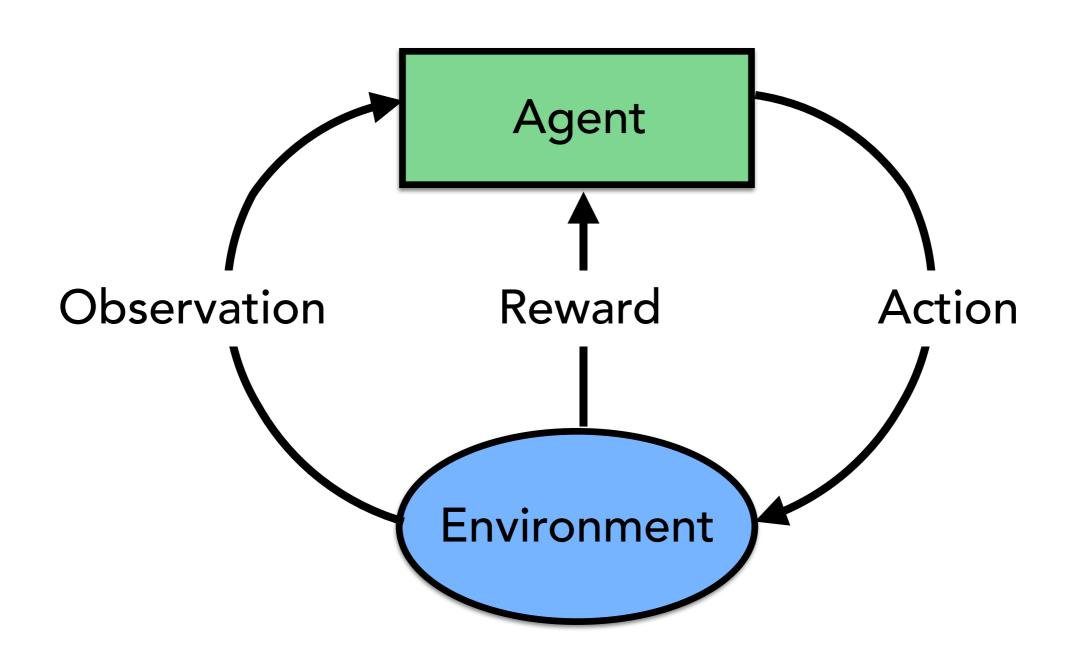
$$L_i(t) = L_i^0 + L_i^1 t + \dots$$



TDIR with NN



Reinforcement learning



Policy gradient

$$(a_0, r_0, a_1, r_1, \dots)$$
 Action - reward pairs, trajectory

$$R(au) = \sum_{t=0}^{T} \gamma^t r_t$$
 Reward

$$J(\pi) = \mathbf{E}_{\tau \sim \pi}[R(\tau)]$$
 Performance

$$\nabla_{\Theta} J(\pi) = \mathbf{E}[\nabla_{\Theta} \log \pi(a_t | s_t) R(\tau)]$$

Gradient of performance can be approximated by sampling