

Contribution

We present a **new algorithm** for estimating the relative translation and orientation between an inertial measurement unit (IMU) and a camera, that are rigidly connected (see photo below).

Motivation

- The **combination** of vision and inertial sensors is very suitable for many applications.
- The present algorithm provides an **easy to use** algorithm which does not require additional hardware, except a piece of paper with a checkerboard pattern.
- For high precision applications, good sensor calibration is **absolutely essential**.

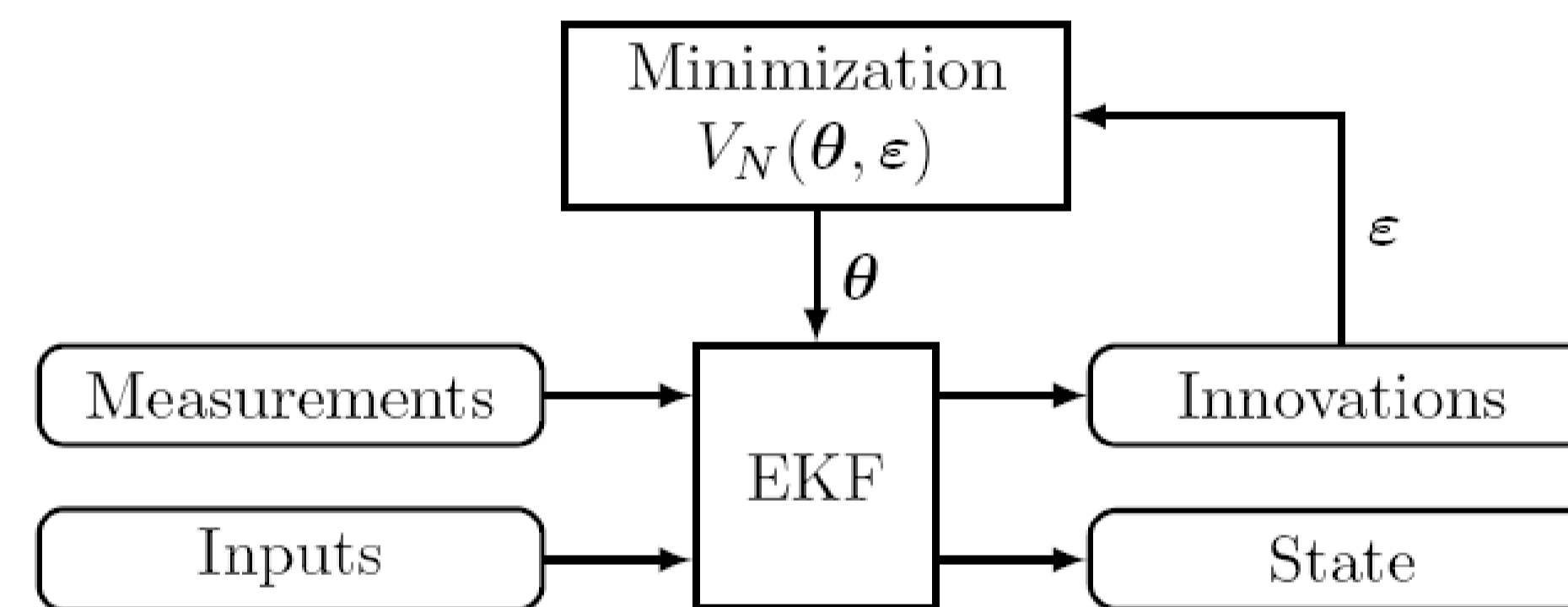


Algorithm

The problem can be cast as a **gray-box** system identification problem. The algorithm is based on a model of the sensor unit motion

$$\begin{aligned} x_{t+1} &= f(x_t, u_t, \theta) + w_t, \\ y_t &= h(x_t, \theta) + e_t, \end{aligned}$$

which of course depends on the relative translation and orientation, here denoted θ .



Based on the model we can form and solve

$$\begin{aligned} \hat{\theta} &= \arg \min_{\theta} V(\theta, Z), \\ V(\theta, Z) &= \frac{1}{2} \sum_{t=1}^N \|\epsilon_t(\theta)\|_{S_t(\theta)^{-1}}^2, \end{aligned}$$

where the prediction error $\epsilon_t(\theta)$ is given by

$$\epsilon_t(\theta) = y_t - \hat{y}_{t|t-1}(\theta)$$

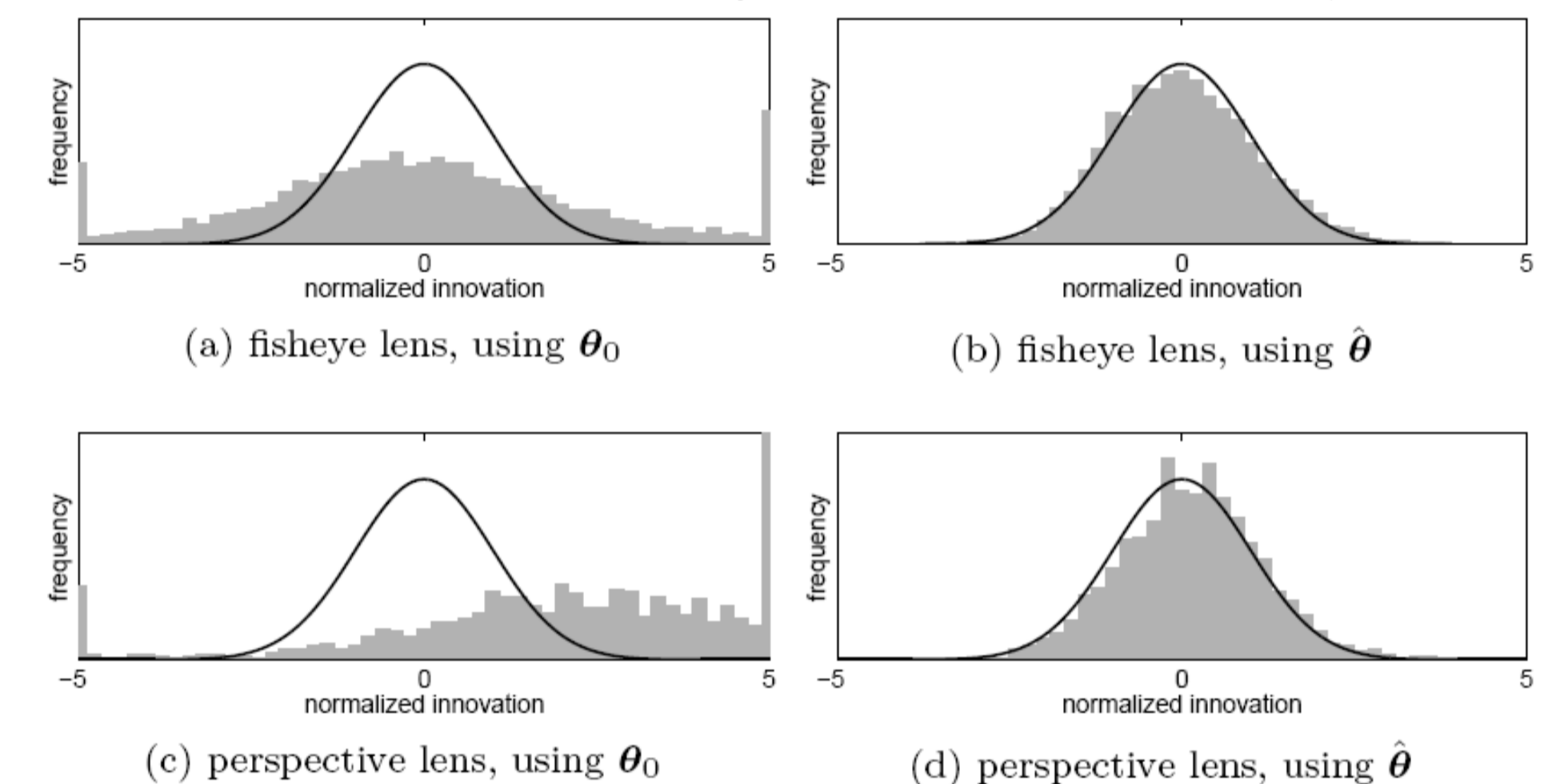
Finally, we make use of an **EKF** to find the predictions.

Experiments

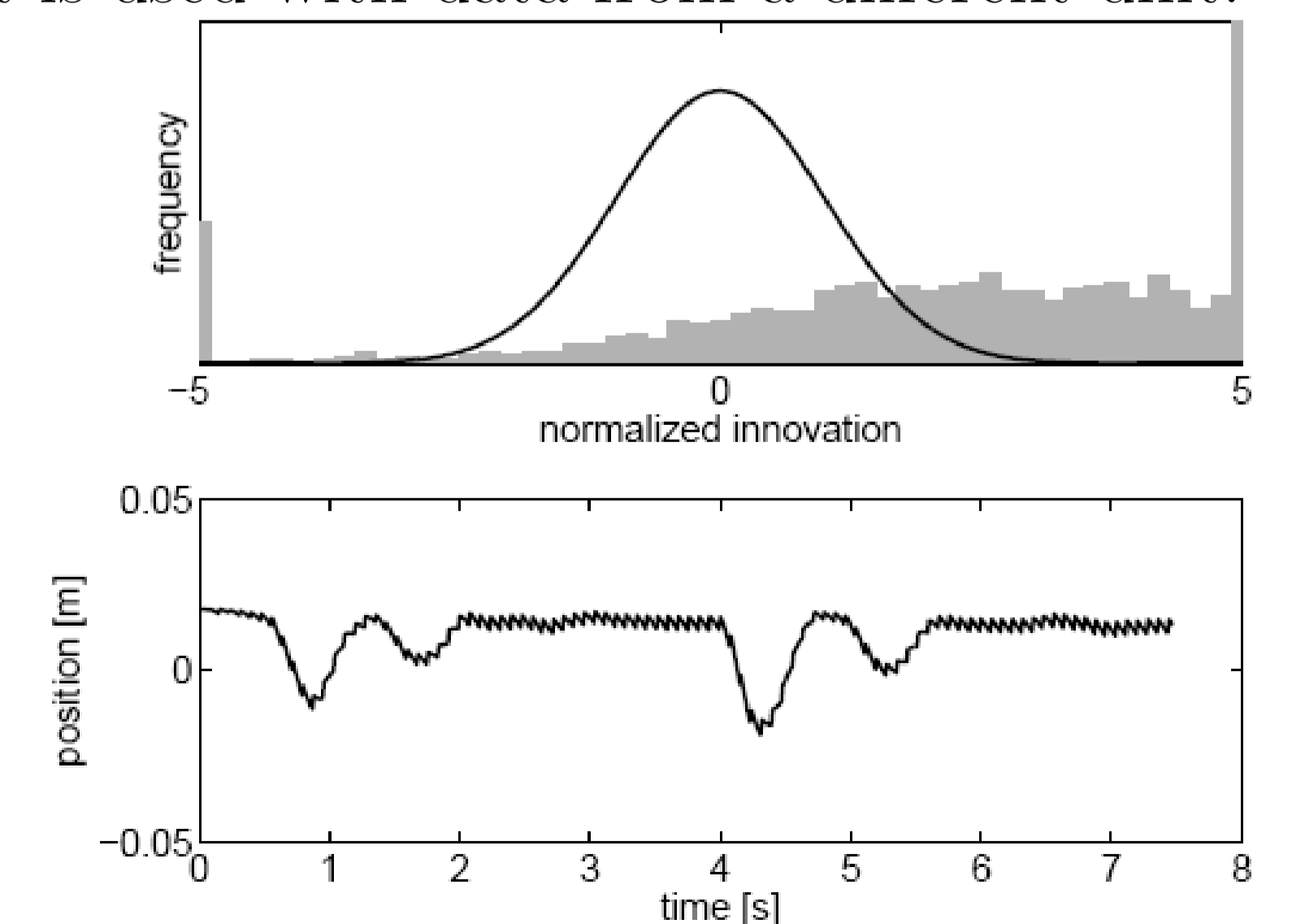
Below we plot the **normalized innovations**

$$\epsilon_t = S_t^{-1/2} \epsilon_t$$

for validation data using the initial guess and the result from our new algorithm, respectively.



In the plot below the identified θ from one sensor unit is used with data from a different unit.



Clearly accurate calibration is **necessary**.