



# Joint Navigation and Synchronization using SOOP in GPS-denied environments: Algorithm and Empirical Study

Leng Mei

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# Outline

- 1 Introduction
- 2 Measurements
  - measurement type
  - measurement model
- 3 Algorithm
- 4 Experiment
  - experiment setup
  - results
- 5 Conclusion

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# Motivation

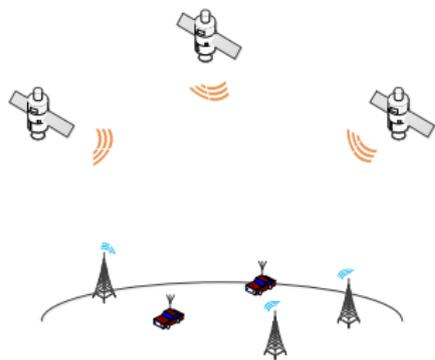


Figure: environment under investigation

- GPS-denied:
  - ▶ requires dedicated GPS receiver
  - ▶ requires open sky view
  - ▶ weak GPS signal close to noise level
  - ▶ deliberately disabled by adversaries
- signal-of-opportunity (SOOP):
  - ▶ widely available from existing infrastructure
  - ▶ relatively high SNR
  - ▶ requires *certain* prior knowledge
  - ▶ synchronization issue
    - ★ oscillators with poor quality
    - ★ passive beacon

# Goal

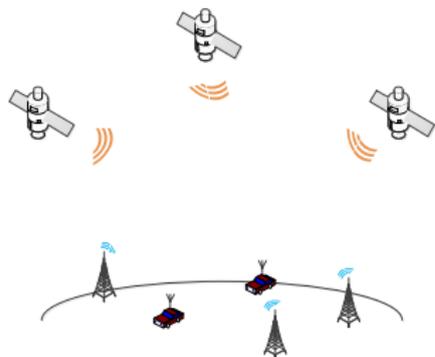


Figure: environment under investigation

Goal: with two cooperative receivers, to jointly track **a target receiver's state** and **its clock drifting** with respect to its cooperative peer.

## Scenarios:

- ▶ two receivers, mobile & unsynchronised
- ▶ two receivers cooperate with each other
- ▶ capture SOOP in an “eavesdropping” way
- ▶ minimum prior knowledge:
  - ★ unknown signal structure
  - ★ unknown transmit time
  - ★ unknown transmit power
  - ★ knowns: beacon states (position, velocity)

# Navigation Scheme

- target receiver:  $G$
- anchor receiver:  $A$
- cooperation:
  - ▶ two receivers see a common set of beacons
  - ▶  $A$  shares with  $G$  its received signal or beacon information derived from its received signal
  - ▶  $A$  shares with  $G$  its own state information

# Navigation Scheme

- 1 **spectrum scanning**: identify near-by available beacons
  - 2 **handshaking**: agree on the beacon to receive from and the time to receive
  - 3 **information exchanging**:  $A$  shared its received information and its state information with  $G$
  - 4 **tracking**:  $G$  perform self-localization and self-synchronization with information from  $A$  and its own received signal.
- SOOP are ad hoc
  - bursts from beacons are received in a sequential way

# Questions to answer

## ① Measurements

- ▶ what types of measurements to use?
- ▶ how does clock drifting affect the measurement? → proper measurement model

## ② Tracking algorithm:

- ▶ can we adapt extended Kalman filter for this problem?
- ▶ how does the nonlinearity and uncertainty affect the performance?

## ③ Field experiment:

- ▶ how well does our measurement model fit in real life?
- ▶ how well does our algorithm perform in real life?

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# Measurement Type

- RSS? TOA? AOA?
- TDOA/FDOA ← unknown waveform & signal characteristics
- methods to obtain measurements: cross-correlation

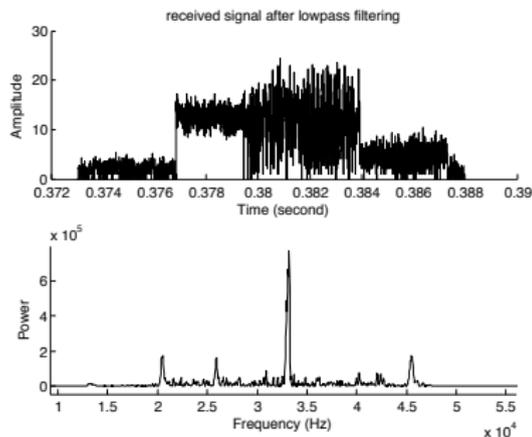


Figure: received packets from Iridium

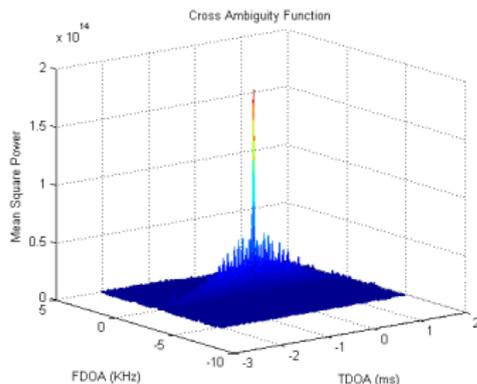


Figure: complex ambiguity function between raw signals

# Measurement Model

at the  $l$ -th time slot

$$\tau_b^{(l)} \approx \left\| \mathbf{p}_1^{(l)} - \mathbf{s}_b^{(l)} \right\| - \left\| \mathbf{p}_2^{(l)} - \mathbf{s}_b^{(l)} \right\| + T_e^{(l)} \alpha^{(l)} + \theta^{(l)} + \varpi_b^\tau, \quad (1a)$$

$$\xi_b^{(l)} \approx (\mathbf{v}_1^{(l)} - \mathbf{v}_b^{(l)})^T \mathbf{u}_{1,b}^{(l)} - (\mathbf{v}_2^{(l)} - \mathbf{v}_b^{(l)})^T \mathbf{u}_{2,b}^{(l)} + \alpha^{(l)} + \varpi_b^\xi. \quad (1b)$$

- TDOA/FDOA measurement :  $[\tau_b^{(l)}, \xi_b^{(l)}]$ 
  - ▶  $\theta^{(l)}$ : clock offset up to the  $l$ -th time slot  
 $\rightarrow \theta^{(l)} \approx \theta^{(l-1)} + (t_b^{(l)} - t_b^{(l-1)})\alpha^{(l-1)}$
  - ▶  $\alpha^{(l)}$ :  $c(\beta_2^{(l)} - \beta_1^{(l)})$  approximately
  - ▶  $\varpi_b^\tau$  and  $\varpi_b^\xi$ : measurement noise, assumed Gaussian.
- knowns:  $\mathbf{s}_b^{(l)}, \mathbf{v}_b^{(l)}, \mathbf{p}_2^{(l)}, \mathbf{v}_2^{(l)}$
- unknowns:  $[\mathbf{p}_1^{(l)}, \mathbf{v}_1^{(l)}, \theta^{(l)}, \alpha^{(l)}]$

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# Dynamic Model

$$\Delta_l \triangleq t_b^{(l)} - t_b^{(l-1)}$$

- clock:

$$\begin{bmatrix} \theta^{(l)} \\ \alpha^{(l)} \end{bmatrix} = \begin{bmatrix} 1 & \Delta_l \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \theta^{(l-1)} \\ \alpha^{(l-1)} \end{bmatrix} + \boldsymbol{\nu}_c^{(l)}, \quad (2)$$

- receiver movement:

$$\begin{bmatrix} \mathbf{p}^{(l)} \\ \mathbf{v}^{(l)} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \Delta_l & \Delta_l & \Delta_l \\ 0 & 0 & 0 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{p}^{(l-1)} \\ \mathbf{v}^{(l-1)} \end{bmatrix} + \boldsymbol{\nu}_s^{(l-1)}, \quad (3)$$

- important parameters: covariance of  $\boldsymbol{\nu}_c^{(l)}$  and  $\boldsymbol{\nu}_s^{(l)}$ 
  - ▶  $\boldsymbol{\nu}_c^{(l)}$ : depends on intensity of the diffusion process of clock components.
  - ▶  $\boldsymbol{\nu}_s^{(l)}$ : depends on acceleration.

# Extended Kalman Filter

$$\text{dynamics: } \mathbf{x}^{(l)} = \mathbf{H}_l \mathbf{x}^{(l-1)} + \boldsymbol{\nu}^{(l)}, \quad (4)$$

$$\text{measurements: } \mathbf{y}^{(l)} = f(\mathbf{x}^{(l)}) + \boldsymbol{\varpi}^{(l)}. \quad (5)$$

- Gaussian assumption:
  - ▶  $\boldsymbol{\nu}^{(l)} \sim \mathcal{N}(0, \mathbf{Q}^{(l)})$ : prior knowledge
  - ▶  $\boldsymbol{\varpi}^{(l)} \sim \mathcal{N}(0, \mathbf{R}^{(l)})$ : measurement accuracy  $\rightarrow$  CRLB
- linearisation:  $\mathbf{F}_l = \nabla_{\mathbf{x}} f(\mathbf{x})|_{\mathbf{x}=\mathbf{x}^{(l)}}$

$$\mathbf{m}_{l|l-1} = \mathbf{H}_l \mathbf{m}_{l-1|l-1}, \quad (6a)$$

$$\mathbf{P}_{l|l-1} = \mathbf{Q}_l + \mathbf{H}_l \mathbf{P}_{l-1|l-1} \mathbf{H}_l^T, \quad (6b)$$

$$\mathbf{P}_{l|l} = \left( \mathbf{P}_{l|l-1}^{-1} + \mathbf{F}_l^T \mathbf{R}_l^{-1} \mathbf{F}_l \right)^{-1}, \quad (6c)$$

$$\mathbf{m}_{l|l} = \mathbf{m}_{l|l-1} + \mathbf{K}_l (\mathbf{y}^{(l)} - f(\mathbf{m}_{l|l-1})), \quad (6d)$$

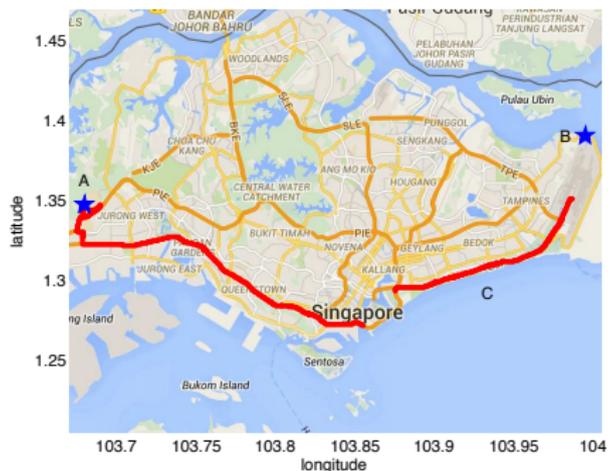
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# Experiment



Figure: one set of receiver



- receiver: USRP N210+WBX
- beacon: Iridium satellites
- off-line processing:
  - ▶ spectrum scanning, handshaking, and information exchanging are correctly carried out.

# TDOA/FDOA measurement

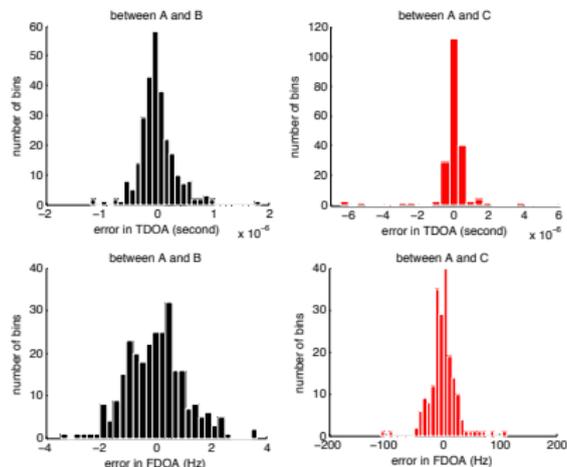


Figure: histogram for TDOA/FDOA estimation error

- mean: close to zero
- standard deviation:
  - ▶ TDOA:  $2.73 \mu\text{s}$  for  $A-B$ ,  $1.73 \mu\text{s}$  for  $A-C$
  - ▶ FDOA:  $2.19 \text{ Hz}$  for  $A-B$ ,  $36.8 \text{ Hz}$  for  $A-C$

# Measurement Model Correctness

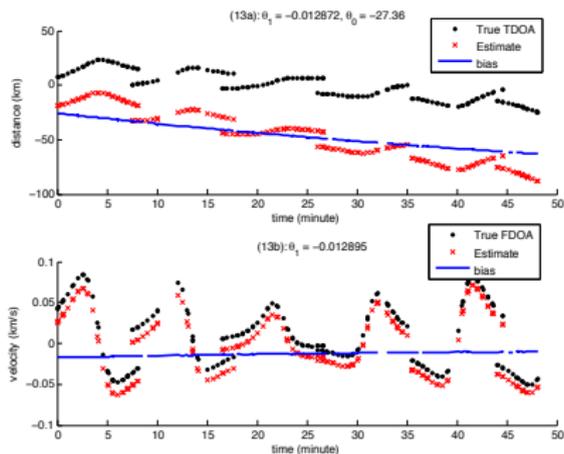
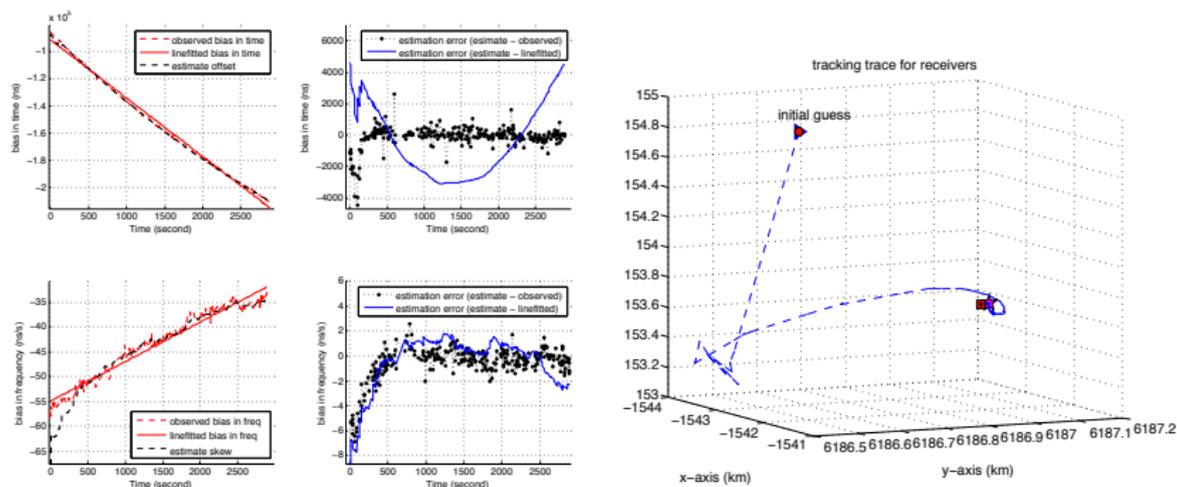


Figure: values of TDOA/FDOA v.s its true value

- from TDOA bias  $\theta^{(l)} = \theta^{(0)} + T\alpha^{(l)}$ : empirical value for  $\alpha$  -0.012872
- from FDOA bias  $\alpha^{(l)}$ : empirical value for  $\alpha$  -0.012895

# Localizing static receiver - 1



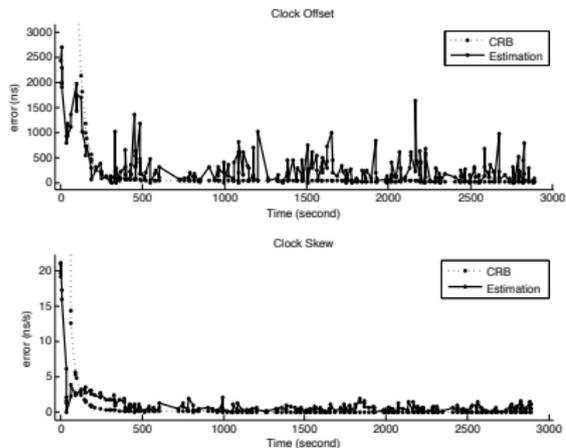
(a) Tracking clock drifting parameters

(b) Trace for estimating the receiver location

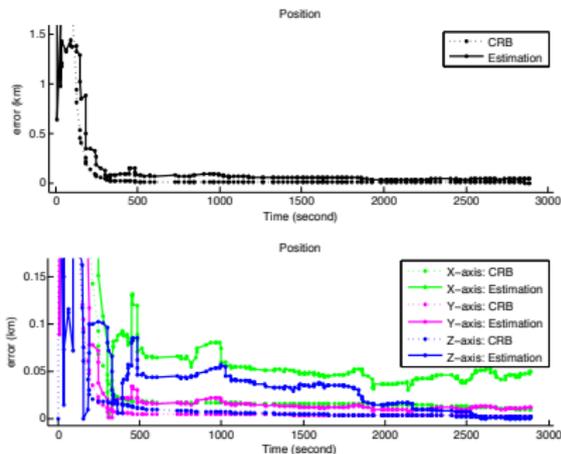
Figure: jointly estimating static receiver  $B$ 's location and clock parameters.

- slightly increasing clock skew

# Localizing static receiver - 2



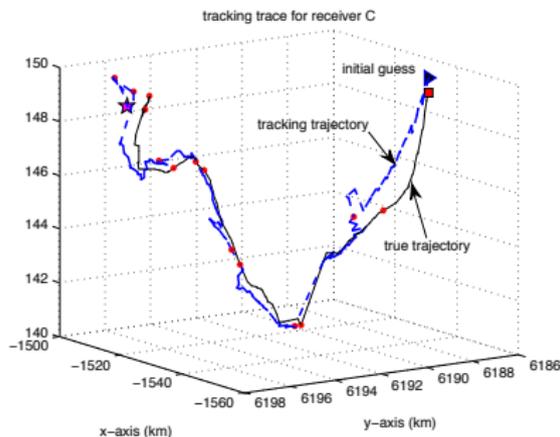
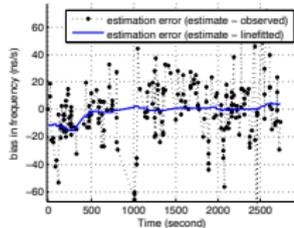
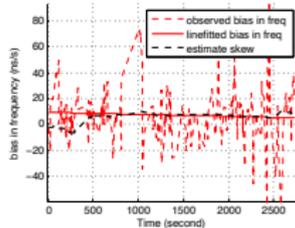
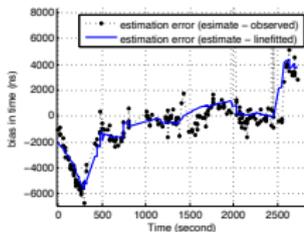
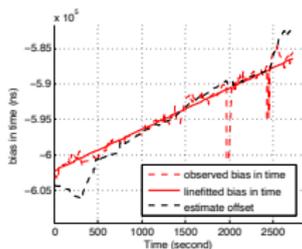
(a) clock parameters



(b) position

- RMSE smaller than 50 m within 5 minutes
- clock skew model mismatch in CRLB

# Tracking manoeuvring receiver



(c) Tracking clock drifting parameters

(d) Tracking the receiver state

**Figure:** The proposed algorithm for tracking the manoeuvring receiver *C*.

# Summary

- ① bias caused by clock drifting:
  - ▶ in TDOA: time-varying and linearly depends on clock skew
  - ▶ in FDOA: device dependent, roughly constant or slowly time-varying
- ② sequential tracking algorithm:
  - ▶ dynamic model matters:
    - ★ correctly tracking the state and the clock parameters when the model fits well
    - ★ problematic when tracking manoeuvring receiver with insufficient measurements or incorrect model information
  - ▶ initial guess matters:
    - ★ the clock biases and the receiver states are correlated.

# Future Work

- incorporate IMU to improve accuracy for tracking manoeuvring target.
  - ▶ to compensate for insufficient measurements due to long observation intervals
  - ▶ must deal with the accumulating error in IMU
- extend to scenarios with multiple targets.
- explore alternative beacons, including UAV, planes, and FM stations.

Thank you!