

Variational Message Passing-based Respiratory Motion Estimation and Detection Using Radar Signals

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Abstract

- Detect **infants in cars** using UWB radar.
- Considering **multipath** propagation and clutter.
- Use **respiratory motion** to differentiate the target from the clutter.
- **Approximate model evidence** for the detection.

Conclusion

- **Multipath information boosts SNR!**
- **Channel, respiratory motion and model evidence** estimated by variational message passing (VMP).

Signal Model

Received signal can be modeled as outer product of respiratory motion and channel

$$\mathbf{R} = \mathbf{h}\mathbf{b}^T + \text{AWGN} + \text{clutter}, \quad \mathbf{R} \in \mathbb{C}^{N \times M}.$$

- **Channels** \mathbf{h} are modeled as independent Gaussian processes and cover the multipath propagation inside the car.
- **Respiratory motion** \mathbf{b} is also modeled as Gaussian process.
- Additive white Gaussian **noise** (AWGN) with unknown precision λ .
- **Clutter** is assumed static and removed by subtracting the mean over slow time.

Detection

Bayes: compare model evidence

$$p(\text{occupied}) \leq p(\text{empty}).$$

Obtain $p(\text{occupied})$ by integrating out \mathbf{b} , \mathbf{h} and the (unknown) noise precision λ

$$p(\text{occupied}) \propto \int \underbrace{p(\mathbf{R}|\mathbf{b}, \mathbf{h}, \lambda) \cdot p(\mathbf{b})p(\mathbf{h})p(\lambda)}_{\text{posterior } p(\mathbf{b}, \mathbf{h}, \lambda|\mathbf{R})} d\mathbf{b} d\mathbf{h} d\lambda.$$

Solving the integral is not feasible!

Solution: approximate posterior with simpler distribution to make integral tractable \rightarrow VMP.

- $p(\mathbf{b}, \mathbf{h}, \lambda|\mathbf{R}) \approx q(\mathbf{b}, \mathbf{h}, \lambda)$
- $p(\text{occupied}) \approx \mathcal{L}(q)$

Message Passing

Maximize the evidence lower bound (ELBO) $\mathcal{L}(q)$ to approximate evidence

$$p(\text{occupied}) = \mathcal{L}(q) + \mathcal{D}_{\text{KL}}(q||p)$$

based on the approximating distribution

$$q(\mathbf{b}, \mathbf{h}, \lambda) = q_{\mathbf{b}}(\mathbf{b}) \cdot q_{\mathbf{h}}(\mathbf{h}) \cdot q_{\lambda}(\lambda).$$

Using conjugate priors results in distributions from well known families:

- $q_{\mathbf{b}}(\mathbf{b}) = \mathcal{N}(\mathbf{b}; \hat{\mathbf{b}}, \hat{\mathbf{C}}_{\mathbf{b}})$
- $q_{\mathbf{h}}(\mathbf{h}) = \mathcal{CN}(\mathbf{h}; \hat{\mathbf{h}}, \hat{\mathbf{C}}_{\mathbf{h}})$
- $q_{\lambda}(\lambda) = \text{Ga}(\lambda; NM, NM/\hat{\lambda})$

Only update of parameters $\hat{\mathbf{b}}$, $\hat{\mathbf{C}}_{\mathbf{b}}$, $\hat{\mathbf{h}}$, $\hat{\mathbf{C}}_{\mathbf{h}}$ and $\hat{\lambda}$ needed in each iteration.

Illustrations

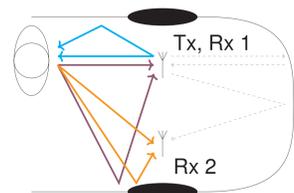
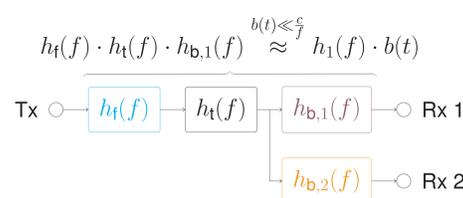
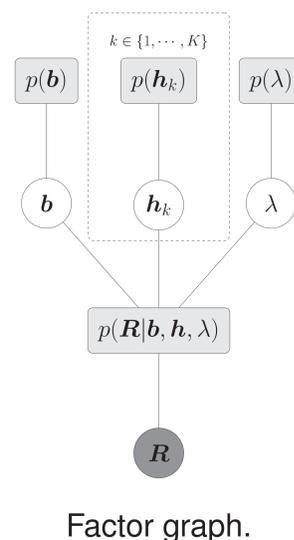


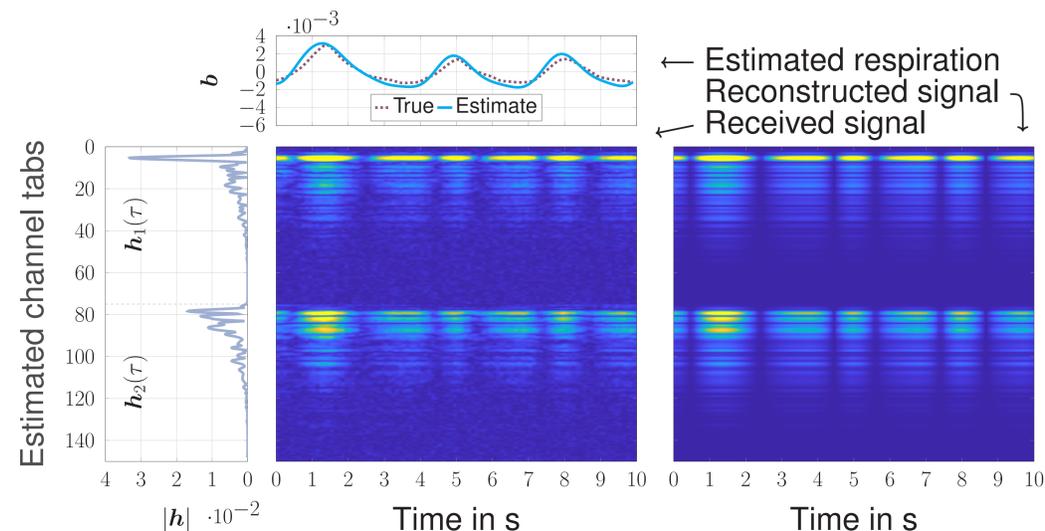
Illustration of the radar setup.



Pinhole channel model.



Factor graph.



Radar measurement example (high SNR).

Results

- Measurement in 2 cars with 34 persons.
- 2 minutes per person, 10s frames.

Comparison approaches:

- FFT: maximum of FFT over slow time.
- Estimator-correlator (EC): based on factorized covariance $\mathbf{C}_{\mathbf{R}} = \mathbf{C}_{\mathbf{b}} \otimes \mathbf{C}_{\mathbf{h}}$.

