Optimal Waveform Design

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Operator-Theoretic Modeling for Radar in the Presence of Doppler

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Introduction Components of Radar

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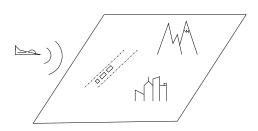
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- Components: transmitters, receivers, scene (targets and clutter), noise (low SNR)
- Under typical assumptions, range encoded as time delay and range rate encoded in frequency shift (narrowband approximation of Doppler)
- The radar chooses the transmitted waveforms and beamforms and the corresponding receiver processing

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- In a SISO radar, the scene acts on the transmitted waveform and produces the received waveform
- For typical scenarios, this transformation is well modeled as a linear operator on $L^2(\mathbb{R})$ plus noise; target and clutter operators summed
- The design objective varies with application goal (detection, classification, scene characterization)

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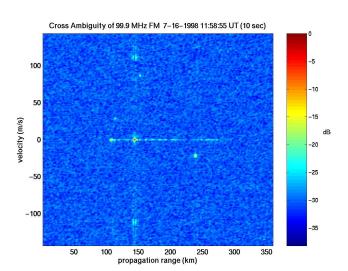
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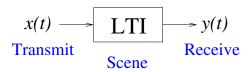
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- Modeling of a radar scene as a linear time-invariant (LTI) system is common in the radar literature
- In particular, LTI models have received attention in connection with waveform design
- But LTI models have a fundamental limitation...

Introduction Limitations of LTI models

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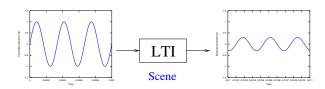
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- Sinusoids are eigenfunctions of LTI systems
- So the only frequency components present in the output are those present in the input
- LTI models cannot accommodate Doppler!

Hilbert-Schmidt Operators

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Our proposal: Replace LTI operator scene models with Hilbert-Schmidt operator scene models, which accommodate Doppler and have other desirable features

• Fundamental building block: displacement operator $D(p,q):L^2(\mathbb{R})\to L^2(\mathbb{R})$ by

$$D(p,q)f(t) = e^{ipt}f(t-q)$$

Here p denotes frequency shift and q time shift; the (p,q) notation is adopted from physics where they are standard for phase space coordinates

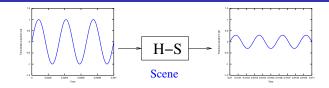
• Every Hilbert-Schmidt class operator $S:L^2(\mathbb{R})\to L^2(\mathbb{R})$ can be written as a superposition of displacement operators

$$S = \int_{\mathcal{D}} \int_{\mathcal{Q}} s(p,q) D(p,q) \, dp \, dq$$

Properties of Hilbert-Schmidt Operators

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- H-S operators can impart frequency shifts as well as time and phase shifts
- So HS models can accommodate Doppler
- LTI (convolutional) operators are a subclass of the H-S class
- But...

$$D(p,q)D(p',q') = D(p',q')D(p,q)e^{i(pq'-p'q)}$$

i.e., H-S operators are generally non-commutative

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• Discrete target with position (p_0, q_0) :

$$\sum_{(p,q)\in N(0,0)} s_t(p,q)D(p,q) D(p_0,q_0)$$

• Clutter:

$$\sum_{(p,q)} s_c(p,q) D(p,q)$$

e.g., Markov random field, homogeneous, Gaussian...

- Both target and clutter are often sparse, but not in a finite library
- *Noise:* Additive Gaussian n(t)

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Standard radar set-up...

- Transmitter: Baseband waveform w(t) with modulation to carrier frequency p_0
- Receiver: Demodulation plus signal processing (H)

Typical goal:

Define optimal systems (waveform w & processing H) for various classes of problems, ultimately with *constraints* (e.g., power, bandwidth, time) on w

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Receiver data at baseband...

Target present (H₁)

$$x(t) = \sum_{(p,q) \in N(0,0)} s_t(p,q) D(p_0,q_0)^{\dagger}$$

$$D(p,q)D(p_0,q_0)w(t) + n(t)$$

$$= Bw(t) + n(t) \sim \mathcal{N}(Bw(t), \sigma^2)$$

where $D(p_0,q_0)^{\dagger}=\exp(ip_0q_0)D(-p_0,-q_0)$ is the adjoint of $D(p_0,q_0)$

• Target absent (H₀)

$$x(t) = n(t) \sim \mathcal{N}(0, \sigma^2)$$

Waveform optimization

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Define

$$B = \sum_{(p,q)\in N(0,0)} s_t(p,q)D(p_0,q_0)^{\dagger}D(p,q)D(p_0,q_0)$$

• Detection is optimized for fixed ||w|| if $||Bw||^2$ is maximized; i.e., w should maximize

$$\langle Bw, Bw \rangle = \langle w, B^{\dagger}Bw \rangle$$

- and should thus be an eigenfunction of the non-negative definite operator $B^\dagger B$ corresponding to its maximal eigenvalue
- If w is constrained by power rather than energy, the solution changes (another day...)



Scene Model With Clutter

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Summary

Signal at the receiver

$$s(t) = Tw(t) + Cw(t) + n(t)$$

- T is a H-S class target operator
- C is a H-S class clutter operator
- n is white Gaussian noise
- Receiver processes s with a H-S operator H
- Detection decision on the basis of the statistic¹

$$r(t_0) = Hs(t_0) = (HTw + HCw + Hn)(t_0)$$

¹See S. U. Pillai et al., "Waveform design optimization using bandwidth and energy considerations," *Proceedings of the IEEE Radar Conference*, pp. 1–5, May 2008.

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 The associated signal to interference plus noise ratio (SINR) is

$$SINR(t_0) = \frac{|HTw(t_0)|^2}{E|HCw(t_0) + Hn(t_0)|^2}$$
 (1)

- The initial objective is to maximize $\mathsf{SINR}(t_0)$ for a given waveform w
 - ullet The problem of optimizing over w comes up later

Noise Term

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• H may be represented as an integral operator with kernel $\Phi: \mathbb{R}^2 \to \mathbb{C}$

$$Hf(t) = \int_{\mathbb{R}} \Phi(t, \tau) f(\tau) d\tau$$

By direct calculation,

$$\mathsf{E}|Hn(t)|^2 = \sigma^2 \int |\Phi(t,\tau)|^2 d\tau$$

• Denoting $h_{t_0} = \Phi(t_0, \cdot)$, this gives

$$\mathsf{E}|Hn(t_0)|^2 = \sigma^2||h_{t_0}||^2 \tag{2}$$

Basic Detection Problem Clutter Term

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For the clutter term,

$$|HCw(t_0)|^2 = \int \int h_{t_0}(\tau) \overline{h_{t_0}(u)} Cw(\tau) \overline{Cw(u)} d\tau du$$
$$= \langle h_{t_0}, G_C h_{t_0} \rangle$$
(3)

• $G_C \ge 0$ is a "waveform dependent" Hermitian operator defined by

$$G_C f(t) = \int_{\mathbb{D}} \mathsf{E}[Cw(t)\overline{Cw(\tau)}] f(\tau) d\tau$$

- Note that the clutter operator may be regarded as random; no issues if n has zero mean and is independent of C
- For deterministic (known) clutter

$$G_C = CP_w C^{\dagger}$$

where P_w is projection onto span $w_{\text{total}} = 0.00$

Signal Term and Maximization Formulation

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• In inner product form, the SINR numerator is

$$|HTw(t_0)|^2 = \langle h_{t_0}, P_{Tw}h_{t_0} \rangle$$
 (4)

- ullet P_{Tw} is the rank-one projection operator onto span Tw
- Substituting (2), (3), and (4) into (1) yields

$$SINR(t_0) = \frac{\langle h_{t_0}, P_{Tw} h_{t_0} \rangle}{\langle h_{t_0}, (\sigma^2 \mathbb{I} + G_C) h_{t_0} \rangle}$$
 (5)

- SINR maximization is thus a generalized eigenvalue problem
 - $P_{Tw} > 0$ with rank one
 - $G_C \ge 0$, so $(\sigma^2 \mathbb{I} + G_C) > 0$ provided $\sigma^2 > 0$
- Hence SINR is maximized by setting

$$h_{t_0} = h_{\text{max}} = (\sigma^2 \mathbb{I} + G_C)^{-1/2} Tw$$

Basic Detection Problem Maximizing SINR

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• Denote $G = (\sigma^2 \mathbb{I} + G_C) > 0$ and observe G is self-adjoint

• Then $h_{\text{max}} = G^{-1/2}Tw$

$$\mathrm{SINR}_{\mathrm{max}} = \frac{\left\langle G^{-1/2}Tw, P_{Tw}G^{-1/2}Tw \right\rangle}{\left\langle G^{-1/2}Tw, G^{1/2}Tw \right\rangle}$$

• With f = Tw,

$$\begin{aligned} \mathsf{SINR}_{\mathrm{max}} &= \frac{\left\langle G^{-1/2}f, P_{Tw}G^{-1/2}f \right\rangle}{\left\langle G^{-1/2}f, G^{1/2}f \right\rangle} \\ &= \frac{\left\langle f, G^{-1/2}P_{Tw}G^{-1/2}f \right\rangle}{\left\langle f, f \right\rangle} \end{aligned}$$

Maximizing SINR (cont'd)

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- SINR_{max} is the maximal eigenvalue of $G^{-1/2}P_{Tw}G^{-1/2} > 0$
 - Rank-one ⇒ its trace is its one positive eigenvalue
- Hence

$$\begin{aligned} \mathsf{SINR}_{\mathrm{max}} &= & \mathrm{Tr}(G^{-1/2}P_{Tw}G^{-1/2}) \\ &= & \left\langle G^{-1/2}f, G^{-1/2}f \right\rangle \\ &= & \left\langle f, G^{-1}f \right\rangle = \left\langle Tw, G^{-1}Tw \right\rangle \\ &= & \left\langle Tw, (\sigma^2\mathbb{I} + G_C)^{-1}Tw \right\rangle \end{aligned}$$

This reduces to past results for LTI scene models

Waveform Optimization

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• With deterministic clutter operator C, the problem of finding a waveform w_{\max} to maximize the SINR is

$$w_{\text{max}} = \arg\max\left\langle Tw, (\sigma^2 \mathbb{I} + CP_w C^{\dagger})^{-1} Tw\right\rangle$$

- \bullet The presence of P_w in the denominator makes this problem "non-linear"
- We are investigating an iterative algorithm
 - Take w_0 to be an eigenvector corresponding to the largest eigenvalue of $T^{\dagger}T$ ("target-driven")
 - 2 For k > 0, solve the "linear" problem

$$w_1 = \arg\max \langle Tw, (\sigma^2 \mathbb{I} + CP_{w_{k-1}}C^{\dagger})^{-1}Tw \rangle$$

"Clutter-driven" perspective leads to the same idea



Determining the Scene

Can we get the scene operator from a few measurements?

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Summar

- The preceding examples assume the scene operator is known or hypothetical scene operators are being tested
- Scene operators are often composed of a sparse superposition of D(p,q) operators
- Radar measures the operator one dimension at a time; time is often limited
- Sparsity assumptions of radar scenes may be helpful in regularizing this problem, but there are issues...
 - Operators are infinite-dimensional
 - Sparsity is not in a finite set, and is not usually well approximated this way
 - The SNR is usually very low

Summary Summary and Ongoing Work

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Summary

- Hilbert-Schmidt class operators may be suitable to model radar scenes with Doppler
- Standard detection and waveform optimization problems can be formulated and solved in this framework
- The problem of determining such operators from small numbers of rank-one measurements has not been treated, at lease in the radar literature
- Sparsity assumptions of radar scenes may be helpful in regularizing this problem