
Distributed Sensing and Inference

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- Who am I?
 - Why am I here?
 - Now that I'm here, what should I do?

My Research Interests

Statistical Signal Processing

modeling, parameter estimation, hypothesis testing

- Radar imaging
 - Tomography for radar
 - Multistatic: multiple transmitters and receivers
 - Physics-based modeling for low-dimensional feature sets
 - Discovery, estimation, evaluation
 - Automatic Target Recognition (ATR)
- Sensor Networks
 - Self-calibration (location; orientation)
 - Multi-modal sensor fusion
 - Communication-constrained distributed inference

Radar Feature Modeling and Estimation

$$E(f, \phi) = \sum_{k=1}^n A_k \underbrace{\left(j \frac{f}{f_c} \right)^{\alpha_k}}_{\text{Frequency Dependence}} \underbrace{e^{-2\pi f \gamma_k \sin \phi} \text{sinc} \left(\frac{2\pi f}{c} L_k \sin(\phi - \phi_k) \right)}_{\text{Aspect Dependence}} \underbrace{\exp \left(-j \frac{4\pi f}{c} [x_k \cos \phi + y_k \sin \phi] \right)}_{\text{Location Dependence}}$$

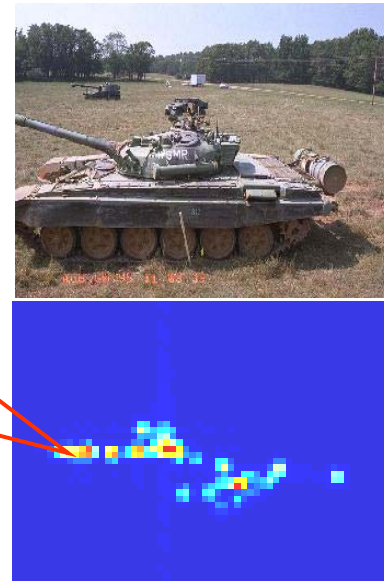
T72-sn132
AZ=249.79, EL=17.18

Scattering Attributes

A_k	=	amplitude [H, V, X]
x_k, y_k	=	location
k	=	frequency type
L_k	=	length
ϕ_k	=	pose angle
α_k	=	angle response

Extract

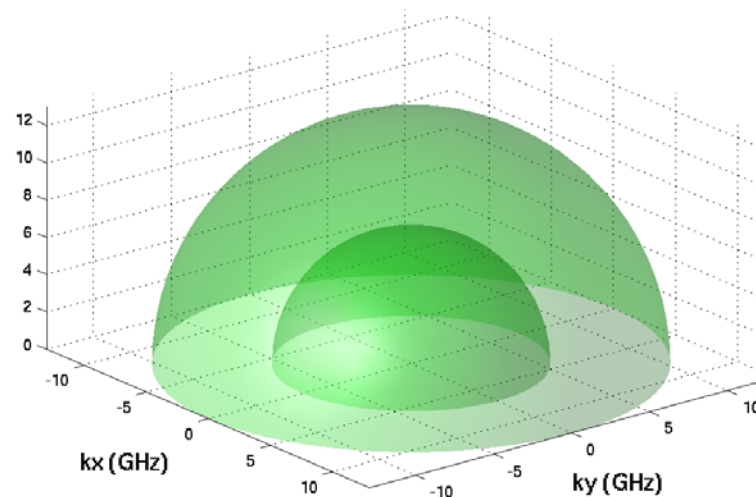
$x=35.2$ m
 $y=46.3$ m
 $A=9.43$ dB
 $\alpha=0.5$
 $L=1.33$ m



Backhoe Data Dome

XpatchF synthetic backscatter data

- Publicly released
- 7-13 GHz frequency band
- 0-360° az, 0-90° elev
- 0.07° angle spacing
- 1" x 1" x 1" resolution
- Full polarization
- ~300 Gbytes of data



3D Radar Tomography Example

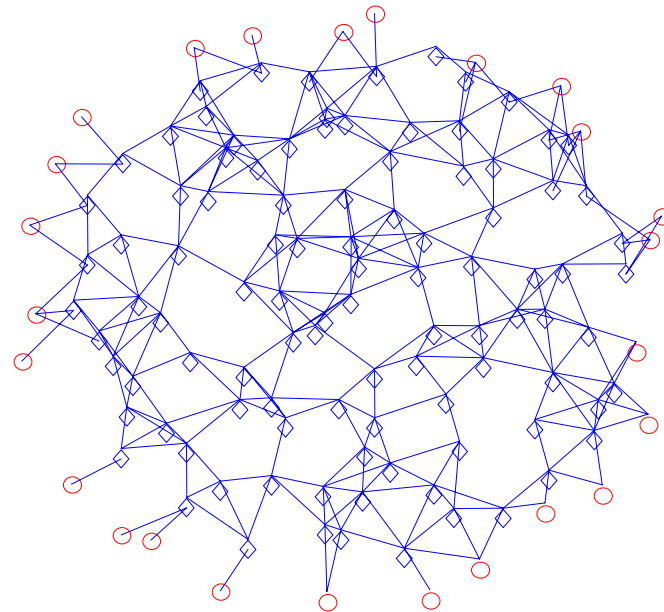
Is there more to signal processing than pretty pictures?



Color encodes
radar look
angle.

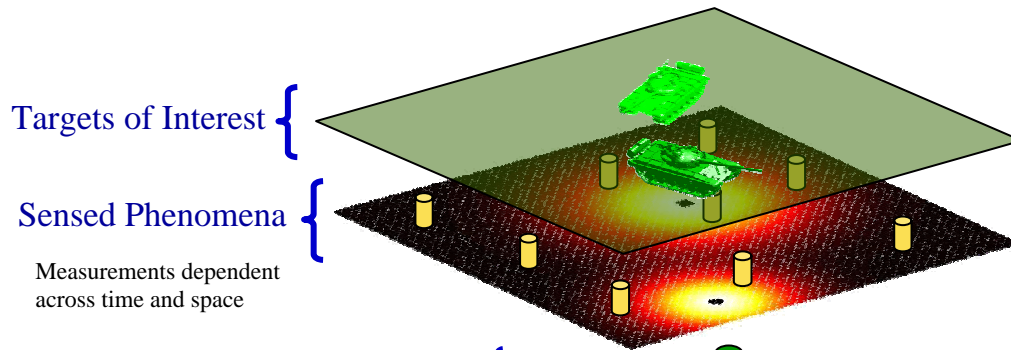
Sensor Network Self-Localization

- Location estimation for nodes in a large ad-hoc network.
- Measurements:
 - Distance
 - Signal time (difference) of arrival
 - Received signal strength
 - Direction of arrival
- Investigate location error as a function of:
 - Node density
 - Number of beacons
 - Network size

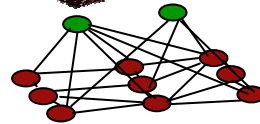


Red: beacon nodes
Blue: nodes w/unknown location

Distributed Inference for Sensor Networks



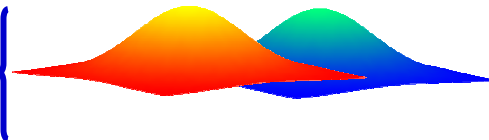
Statistical Model for Sensed Variables and Information Fusion {



Graphical Model for Ad-hoc Communication {



Event Likelihoods for Human Decision-Making {



Collaboration in

Sensing

Communications

Decision-Making by
People

My Research Interests

Statistical Signal Processing

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- Identifiability of models
- Inference quality vs. measurements
 - What measurements do we take to produce ‘best’ estimates or inferences?
- Robust parametric/nonparametric inference
 - Use models where you can
 - Use nonparametric statistical methods on ‘residual’ for robustness
- Automatic inference to support human decisions
 - Explore human machine inference interface
 - SP must generate outputs with higher information rate

Now that I'm here, what should I do?

How do we build SP network 'bipartite graphs'

Epistemology

- [Craig Q2] What can SP tell us about the network?
- [Craig Q3] What should the input signals be?
 - What do we want to know?
 - What can we measure, and why do we think those measurements will be useful?

Jointly define/hypothesize/test a model

model measurements inferences

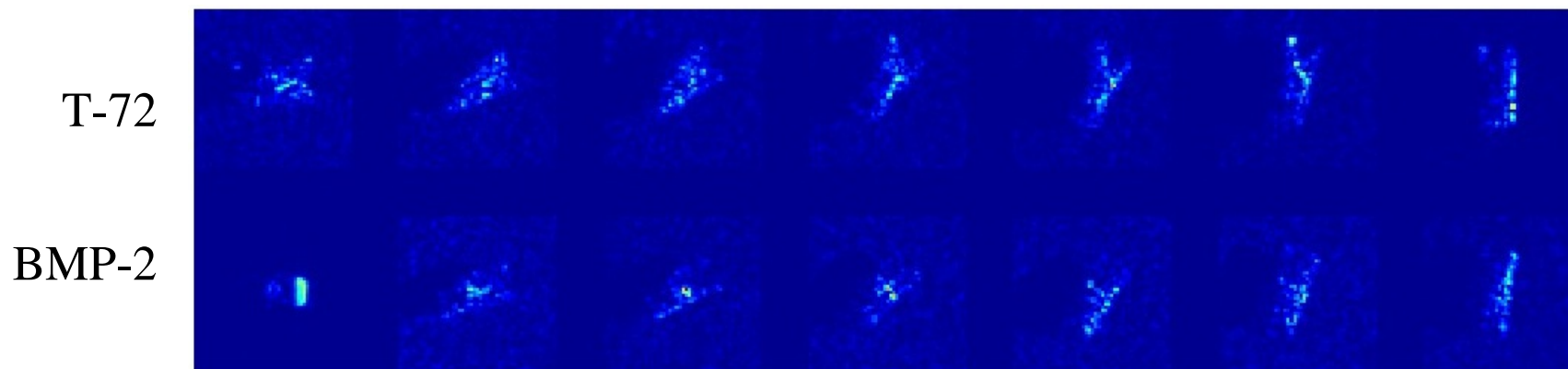
What are the research questions?

- What can SP tell us about the network?
- What's happening in my network?
- How do I detect/identify/quantify (unusual) activity
 - DoS; equipment failures; routing changes
- What can I infer from xxx measurements?
- What other measurements would make the inference task (much) easier?
- How do I present inference outputs to decision-makers?
- How do I automate the inference/decision/response process?

Data Thoughts

- Some high-fidelity data is useful, even if it is synthetic
 - Model development; model robustness; new insights
- Range of data sets: **controlled** **realistic**
 - Identify questions that can/cannot be answered with each set
- Choose measured data sets carefully
 - Goal-oriented
 - Minimize ‘noise’ that detracts from research
- Staged data collections with refinement
- Challenge problems:
 - "If you measure something, it will improve."

ATR: MSTAR Data Set



MSTAR Public Targets:

10 targets x ~275 aspects = 2747 image chips

SP network 'bipartite graphs'

- Problems before tools
- A signal processing 'resource bank'?
 - Triage of research questions?

If you were to design the next internet,
what measurements would you build in?

The answer guides:

- the compelling research questions
- the initial data needs
- a scientific framework for collaboration at the next level