

E1 244: Detection and Estimation

Lecture 1: Introduction



Course information

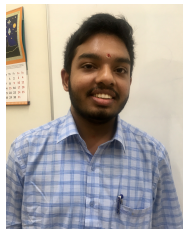
- ▶ Instructor:
 - Sundeep Prabhakar Chepuri.
Email: `spchepuri AT iisc.ac.in`
- ▶ Class schedule:
 - Tuesdays and Thursdays 2.00-3.30pm (Online via MS Teams).
- ▶ Course webpage:
<https://ece.iisc.ac.in/~spchepuri/classes/e1244.html>

Course information

► Teaching Assistants



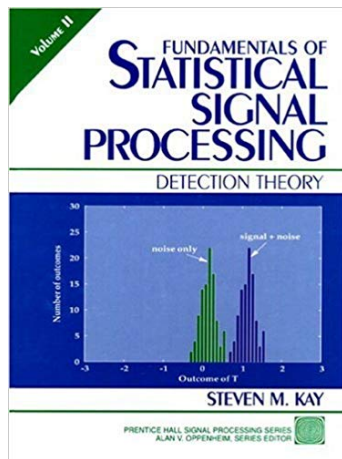
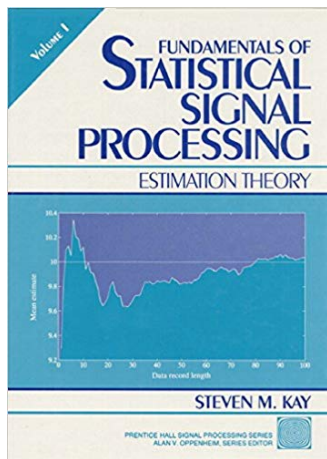
Sravanthi Gurugubelli
sravanthig



Prasobh Sankar
prasobhr

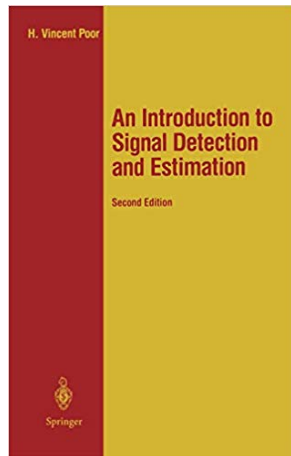
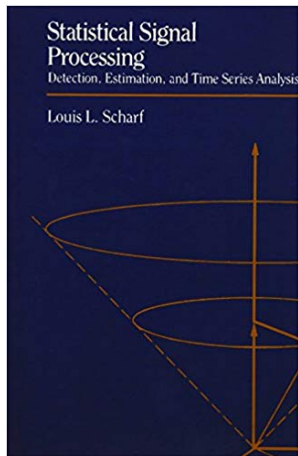
► Exercise/tutorial sessions:

- By TAs, on 1st and 3rd Saturdays 11.00-12.30 pm



- ▶ [Fundamentals of Statistical Signal Processing, Volume I: Estimation Theory](#), S.M. Kay, Prentice Hall 1993, ISBN-13: 978-0133457117.
- ▶ [Fundamentals of Statistical Signal Processing, Volume II: Detection Theory](#), S.M. Kay, Prentice 1993, ISBN-13: 978-0135041352.

Other resources

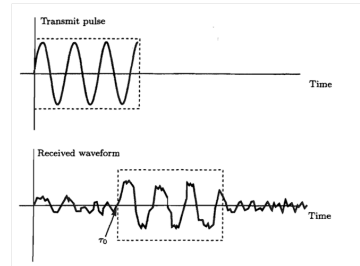
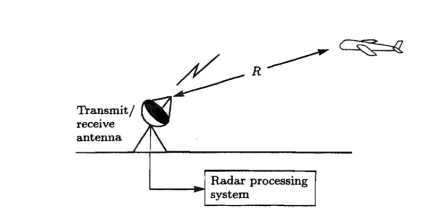


- ▶ [Statistical Signal Processing](#), L.L. Scharf, Pearson India, 2010, ISBN-13: 978-8131733615.
- ▶ [An Introduction to Signal Detection and Estimation](#), H.V. Poor, Springer, 2nd edition, 1998, ISBN-13: 978-0387941738.

Grading and course requirements

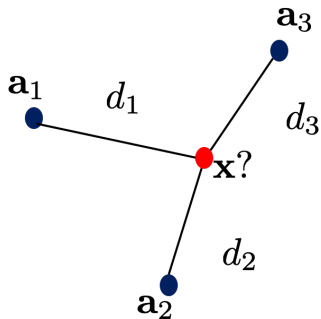
- ▶ **Prerequisite:** Matrix theory (or equivalent) and random processes (or equivalent).
- ▶ **Three assignments** (problem and programming set): 10% each, i.e., 30% in total
- ▶ **Midterm exam:** 20%
 - Open book exam.
- ▶ **Project:** 30%
- ▶ **Final exam:** 20%
 - Open book exam.

Estimation theory



Time delay estimation or ranging

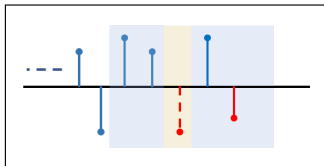
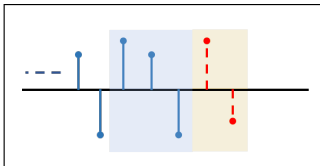
Estimation theory



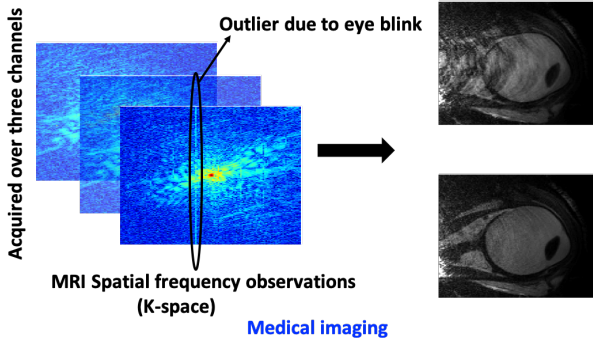
$$d_i = \|\mathbf{x} - \mathbf{a}_i\|_2 + \text{noise}, i = 1, 2, \dots, 3$$

Sensor localization and positioning

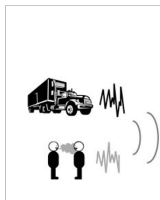
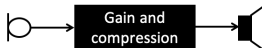
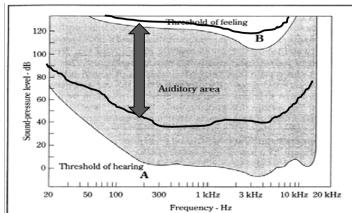
Estimation theory



Time-series prediction and interpolation

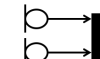


Estimation theory



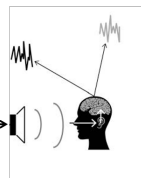
$$y_1 = s + n_1$$

s



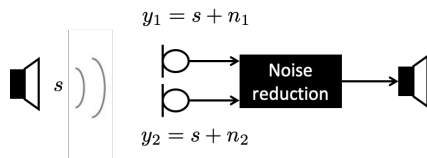
$$y_2 = s + n_2$$

Gain and compression



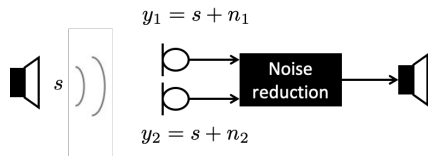
Denosing in hearing aids

A simplified model



How to determine \hat{s} - an estimate of s ?

A simplified model

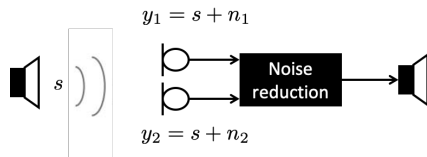


How to determine \hat{s} - an estimate of s ?

- ▶ $\hat{s}_1 = y_1$?
- ▶ $\hat{s}_2 = \frac{1}{N} \sum_{i=1}^N y_i$?

How good are these estimators? Are there better estimators?

A simplified model



How to determine \hat{s} - an estimate of s ?

- ▶ $\hat{s}_1 = y_1$?
- ▶ $\hat{s}_2 = \frac{1}{N} \sum_{i=1}^N y_i$?

Suppose noise at each microphone, n_i , has variance σ_i^2 . Then, how about:

- ▶ $\hat{s}_3 = \frac{\sum_{i=1}^N \frac{y_i}{\sigma_i^2}}{\sum_{i=1}^N \frac{1}{\sigma_i^2}}$?

What is the recipe to determine **optimal** estimators?

Estimation theory

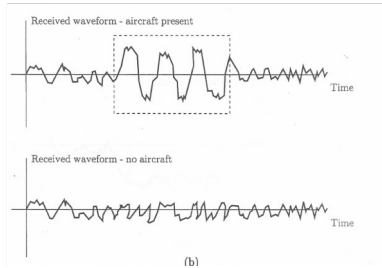
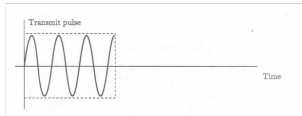
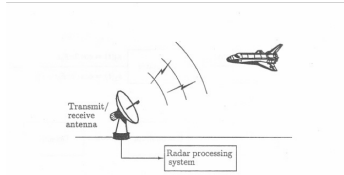
Typical formulation in estimation theory

$$y[n] = f_n(\boldsymbol{\theta}) + w[n].$$

The noise is usually assumed to be stochastic, the parameter vector of interest $\boldsymbol{\theta}$ may be

- ▶ an unknown **deterministic** quantity: classical estimation theory.
- ▶ an unknown **random** quantity: Bayesian estimation theory.

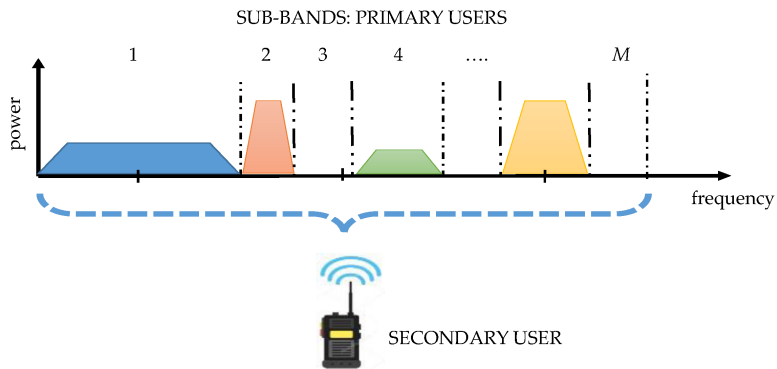
Detection theory



Radar - target detection

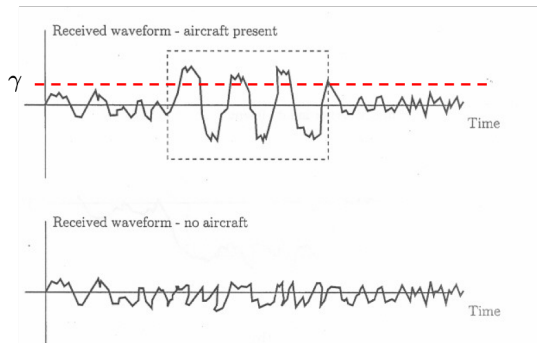
Delay/ranging might not be always required

Detection theory



Cognitive radio - spectrum sensing

Detection theory



How to choose γ ?

Detection theory

Typical detection problem formulation, also referred to as binary hypothesis testing problem:

$$\text{(no target)} \quad \mathcal{H}_0 : x[n] = w[n]$$

$$\text{(target)} \quad \mathcal{H}_1 : x[n] = s[n] + w[n]$$

We wish to infer the **state of nature**, i.e., to decide on \mathcal{H}_0 or \mathcal{H}_1 using a detector of the form:

$$T(x[n]) > \gamma$$

- ▶ How to make an optimal decision: how to choose $T(\cdot)$ and γ ?
- ▶ detection of deterministic signals: Neyman-Pearson detectors.
- ▶ detection with priors: Bayes detectors.
- ▶ detection with unknown parameters.

Content

- ▶ Review of linear algebra and random processes.
- ▶ Minimum variance unbiased estimator
- ▶ Cramér-Rao bound
- ▶ Maximum likelihood estimator
- ▶ Best linear unbiased estimator
- ▶ Least squares and recursive least squares.
- ▶ Structured covariance estimation
- ▶ Bayesian estimators (MMSE and MAP estimators)
- ▶ Kalman filtering
- ▶ Neyman-Pearson detector
- ▶ Bayes detector
- ▶ Multiple hypothesis testing
- ▶ Composite hypothesis testing
- ▶ Sequential probability ratio test (SPRT)

Course objectives

Generally, a solution to estimation and detection problems depend on the underlying **data model** and the **statistical description of the noise and/or unknowns**.

- ▶ How to **mathematically formulate** such problems?
- ▶ Determine optimal estimators, characterize performance of these estimators, and compute the estimation bound.
- ▶ Determine optimal detectors and characterize the performance of these detectors.