

ECE531 Screencast 8.1: Introduction to Detection

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Detection = Hypothesis Testing

Detection is the process of deciding among two or more possible underlying statistical situations (“hypotheses”) from noisy observations.

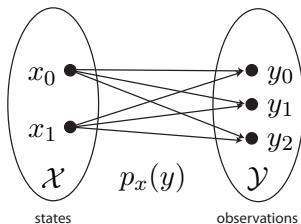
Examples of hypotheses:

- ▶ The coin is fair (\mathcal{H}_0) or not fair (\mathcal{H}_1).
- ▶ The approaching airplane is friendly (\mathcal{H}_0) or unfriendly (\mathcal{H}_1).
- ▶ This email is spam (\mathcal{H}_1) or not spam (\mathcal{H}_0).
- ▶ The medical treatment is effective (\mathcal{H}_1) or ineffective (\mathcal{H}_0).
- ▶ Communication receiver: Given a codebook with M codewords, which codeword was sent ($\{\mathcal{H}_0, \dots, \mathcal{H}_{M-1}\}$)?

More generally, we want to specify a **decision rule** that maps observations to decisions optimally in some sense.

States and Observations

- ▶ Let $x \in \mathcal{X} = \{x_0, \dots, x_{N-1}\}$ denote the **state**, a hidden variable about which we wish to make an inference.
- ▶ The available **observation** is modeled as a random variable Y taking on values in the set \mathcal{Y} .
- ▶ For each state $x \in \mathcal{X}$, we assume we have a **probabilistic description** of the random variable Y when the state is x . The notation $p_x(y) = p_{Y|X}(y|X=x)$ (random X) or $p_x(y) = p_Y(y; x)$ (nonrandom X) is the density of the random variable Y when the state is x .



Hypotheses and Decisions

- ▶ **Hypotheses** can be represented as a **partition** of \mathcal{X} , denoted by $\mathcal{H} = \{\mathcal{H}_0, \mathcal{H}_1, \dots, \mathcal{H}_{M-1}\}$ where

$$\mathcal{H}_i \subseteq \mathcal{X}$$

$$\mathcal{H}_i \neq \emptyset$$

$$\mathcal{H}_i \cap \mathcal{H}_j = \emptyset \text{ for } i \neq j \text{ and}$$

$$\bigcup_i \mathcal{H}_i = \mathcal{X}$$

- ▶ The set of possible **decisions** is then $\mathcal{Z} = \{0, 1, \dots, M - 1\}$ where decision i indicates the selection of hypothesis \mathcal{H}_i . In other words, decision i is the decision that $x \in \mathcal{H}_i$.
- ▶ If \mathcal{X} is finite, then we must have $M \leq N$.

Types of Hypothesis Testing Problems

Recall $N = |\mathcal{X}|$ is the number of states (assume \mathcal{X} is finite for now) and $M = |\mathcal{H}|$ is the number of hypotheses.

- ▶ If $M = 2$, then we have a **binary** hypothesis testing problem.
- ▶ If $M = N$, then we seek to decide the actual state. In this case we can take $\mathcal{H}_i = \{x_i\}$ and we have a **simple** hypothesis testing problem.
- ▶ If $M < N$ or \mathcal{X} is infinite, then we have a **composite** hypothesis testing problem. At least one hypothesis contains more than one state.

Unlike a simple hypothesis with underlying distribution $p_x(y)$, a composite hypothesis does not completely specify the underlying distribution.

Our focus will be on simple hypothesis testing problems for now, but we will return to composite hypothesis testing in a few weeks.

Examples

We have a coin with $\text{Prob}(\text{H}) = q$ unknown.

1. Suppose q can only take on two values: q_0 or q_1 . What kind of hypothesis testing problem is this? **Binary, simple**.
2. Suppose q can take on any value in the set $\{q_0, q_1, \dots, q_{M-1}\}$ and we wish to determine which value it is. What kind of hypothesis testing problem is this? **M -ary, simple**.
3. Suppose q can take on any value in the set $\{q_0, q_1, \dots, q_{N-1}\}$ but only wish to know if it is q_0 or not (e.g. $q_0 = 0.5$ "is the coin fair?"). What kind of hypothesis testing problem is this? **Binary, composite**
 $M = 2 < N$.
4. Suppose q can be any value in $[0, 1]$ and we want to determine this value. What kind of problem is this? **Estimation**.

Model Summary

