

ECE 7251: Signal Detection and Estimation

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Lecture 27, 3/20/02:
Locally Most Powerful Tests

One-Sided LMP Tests

- Usual parametric data model $p(y; \mathbf{q})$
- Consider a single sided problem:

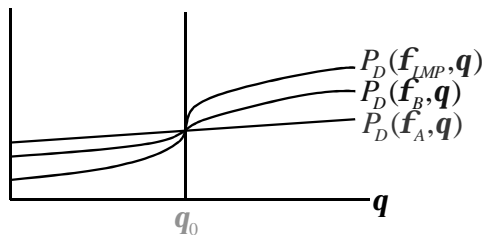
$$H_0 : \mathbf{q} = \mathbf{q}_0, H_1 : \mathbf{q} > \mathbf{q}_0$$

- What to do if there is no UMP test?
- The locally most powerful test of level α has a power curve which maximizes the slope of $P_D(\mathbf{q})$ at $\mathbf{q} = \mathbf{q}_0$

$$\mathbf{f}_{LMP} = \arg \max_{\mathbf{f} \in \{\alpha\text{-level}\}} \frac{d}{d\mathbf{q}} P_D(\mathbf{f}; \mathbf{q}) = \arg \max_{\mathbf{f} \in \{\alpha\text{-level}\}} \frac{d}{d\mathbf{q}} E_{\mathbf{q}}[\mathbf{f}]$$

Graphical Interpretation of LMP Test

$$\alpha = P_{FA}(\mathbf{f}_B) = P_{FA}(\mathbf{f}_A) = P_{FA}(\mathbf{f}_{LMP})$$



Solution to LMP Problem

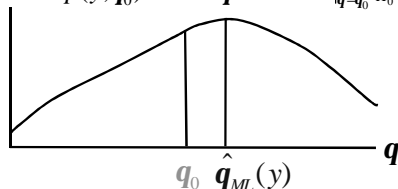
- Using a proof similar to that for the Neyman Pearson lemma, one can show the LMP test is

$$\frac{\left. \frac{d}{d\mathbf{q}} p(y; \mathbf{q}) \right|_{\mathbf{q}=\mathbf{q}_0}}{p(y; \mathbf{q}_0)} \underset{H_0}{\underset{H_1}{\geq}} \mathbf{1} \quad \text{where we pick } \mathbf{1} \text{ to achieve } E_{\mathbf{q}_0}[\mathbf{f}] \leq \alpha$$

- As before, we may need a randomized test if there is a nonzero prob. of landing exactly on the threshold (but we won't worry about that)

Another Interpretation of LMP Test

$$\frac{\left. \frac{d}{d\mathbf{q}} p(y; \mathbf{q}) \right|_{\mathbf{q}=\mathbf{q}_0}}{p(y; \mathbf{q}_0)} = \frac{\left. \frac{d}{d\mathbf{q}} \ln p(y; \mathbf{q}) \right|_{\mathbf{q}=\mathbf{q}_0}}{\ln p(y; \mathbf{q}_0)} \underset{H_0}{\underset{H_1}{\geq}} \mathbf{1}$$



Decide H_0 if the slope at \mathbf{q}_0 is negative or \mathbf{q}_0 is near a stationary point of the loglikelihood, i.e. if \mathbf{q}_0 is near the ML estimate of \mathbf{a}

One-Sided Gaussian Mean Example

$$y_i \in \mathcal{N}(\mathbf{q}, \mathbf{s}^2), H_0 : \mathbf{q} = 0, H_1 : \mathbf{q} > 0$$

$$\frac{d}{d\mathbf{q}} \ln p(y; \mathbf{q}) = \frac{d}{d\mathbf{q}} \left[c - \sum_{i=1}^n \frac{(y_i - \mathbf{q})^2}{2\mathbf{s}^2} \right] \underset{q=0}{\underset{H_0}{\underset{H_1}{\geq}}} \mathbf{1}$$

$$\frac{\bar{ny}}{\mathbf{s}^2} = \sum_{i=1}^n \frac{y_i - \mathbf{q}}{\mathbf{s}^2} \underset{q=0}{\underset{H_0}{\underset{H_1}{\geq}}} \mathbf{1}$$

$$\frac{\sqrt{n}}{\mathbf{s}} \bar{y} \underset{H_0}{\underset{H_1}{\geq}} \mathbf{1} \frac{\mathbf{s}}{\sqrt{n}} = \mathbf{g}$$

Just the UMP test we discussed last lecture

Cauchy Median Example

- Suppose we have n Cauchy samples:

$$p(y; \mathbf{q}) = \prod_{i=1}^n \frac{1}{\pi} \frac{1}{1 + (y_i - \mathbf{q})^2}$$

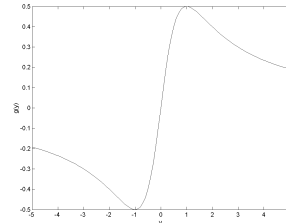
and we want to decide $H_0 : \mathbf{q} = 0, H_1 : \mathbf{q} > 0$

$$\frac{d}{d\mathbf{q}} \ln p(y; \mathbf{q}) = \frac{d}{d\mathbf{q}} \{c - \ln[1 + (y_i - \mathbf{q})^2]\}$$

$$2 \sum_{i=1}^n \frac{(y_i - \mathbf{q})}{1 + (y_i - \mathbf{q})^2} \Big|_{\mathbf{q}=0} \stackrel{H_1}{\geq} 2 \sum_{i=1}^n \frac{y_i}{1 + y_i^2} \stackrel{H_0}{\geq} \mathbf{1}$$

Cauchy Median Example Con't

- Step 1: Pass each data point through a memoryless nonlinearity $g(y) = y/(1 + y^2)$



- Step 2: Sum all the nonlinearity outputs
- Step 3: Compare to a threshold

Two-Sided LMP Tests

- Now consider a double sided problem:

$$H_0 : \mathbf{q} = \mathbf{q}_0, H_1 : \mathbf{q} \neq \mathbf{q}_0$$

- The locally most powerful "unbiased" test of level \mathbf{a} has a power curve which maximizes the *curvature* of $P_D(\mathbf{q})$ at $\mathbf{q} = \mathbf{q}_0$

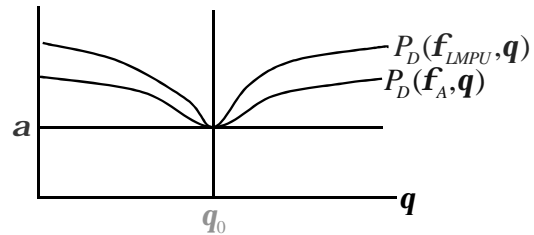
$$\mathbf{f}_{LMP} = \arg \max_{\mathbf{f}} \frac{d^2}{d\mathbf{q}^2} P_D(\mathbf{f}; \mathbf{q}) \Big|_{\mathbf{q}=\mathbf{q}_0}$$

subject to $P_{FA}(\mathbf{f}) = E_{\mathbf{q}_0}[\mathbf{f}] \leq \mathbf{a}$

$$\text{and } \frac{d}{d\mathbf{q}} P_D(\mathbf{f}, \mathbf{q}) \Big|_{\mathbf{q}=\mathbf{q}_0} = \frac{d}{d\mathbf{q}} E_{\mathbf{q}}[\mathbf{f}] \Big|_{\mathbf{q}=\mathbf{q}_0} = 0$$

Graphical Interpretation of LMPU Test

$$\mathbf{a} = P_{FA}(\mathbf{f}_A) = P_{FA}(\mathbf{f}_{LMPU})$$



- Say a test is unbiased if $P_D(\mathbf{q}) > P_{FA}$ for all \mathbf{q}

Solution to LMPU Problem

- Using Lagrange multipliers, one can show that the test is

$$\frac{d^2}{d\mathbf{q}^2} p(y; \mathbf{q}) \Big|_{\mathbf{q}=\mathbf{q}_0} \stackrel{H_1}{\geq} \mathbf{1} \left(p(y; \mathbf{q}_0) + \mathbf{r} \frac{d}{d\mathbf{q}} p(y; \mathbf{q}) \Big|_{\mathbf{q}=\mathbf{q}_0} \right)$$

where $\mathbf{1}$ and \mathbf{r} are picked to satisfy constraints

- If we're really lucky, it sometimes turns out that $\mathbf{r} = 0$
- $$\frac{d^2}{d\mathbf{q}^2} p(y; \mathbf{q}) \Big|_{\mathbf{q}=\mathbf{q}_0} \stackrel{H_1}{\geq} \mathbf{1} \frac{p(y; \mathbf{q}_0)}{H_0}$$

Two-Sided Gaussian Mean Example

$$y_i \in \mathcal{N}(\mathbf{q}, \mathbf{s}^2), H_0 : \mathbf{q} = 0, H_1 : \mathbf{q} \neq 0$$

$$p(y; \mathbf{q}) = \frac{1}{\sqrt{2\pi\mathbf{s}^2/n}} \exp \left[-\sum_{i=1}^n \frac{(\bar{y} - \mathbf{q})^2}{2\mathbf{s}^2/n} \right]$$

$$\frac{d}{d\mathbf{q}} p(y; \mathbf{q}) = \frac{\bar{y} - \mathbf{q}}{\mathbf{s}^2/n} p(y; \mathbf{q})$$

$$\begin{aligned} \frac{d^2}{d\mathbf{q}^2} p(y; \mathbf{q}) &= \frac{\bar{y} - \mathbf{q}}{\mathbf{s}^2/n} \left[\frac{d}{d\mathbf{q}} p(y; \mathbf{q}) \right] - \frac{1}{\mathbf{s}^2/n} p(y; \mathbf{q}) \\ &= \left[\left(\frac{\bar{y} - \mathbf{q}}{\mathbf{s}^2/n} \right)^2 - \frac{1}{\mathbf{s}^2/n} \right] p(y; \mathbf{q}) \end{aligned}$$

Two-Sided Gaussian Example Con't

$$\frac{d^2}{dq^2} p(y; \mathbf{q}) \Big|_{q=q_0} \Big|_{H_0} \geq \mathbf{I} \left(p(y; \mathbf{q}) + \mathbf{r} \frac{d}{dq} p(y; \mathbf{q}) \Big|_{q=q_0} \right)$$

$$\frac{\left[\left(\frac{\bar{y}}{\mathbf{s}^2/n} \right)^2 - \frac{1}{\mathbf{s}^2/n} \right] p(y; 0)}{\left(p(y; 0) + \mathbf{r} \frac{\bar{y}}{\mathbf{s}^2/n} p(y; 0) \Big|_{q=q_0} \right)} \Big|_{H_0} \geq \mathbf{I}$$

That Pesky Roh Lagrange Multiplier

$$\frac{\bar{y}^2 - \mathbf{s}^2/n}{(\mathbf{s}^2/n + \mathbf{r}\bar{y})} \Big|_{H_0} \geq \mathbf{I}(\mathbf{s}^2/n)$$

- Usual first attempt: Wishful thinking! Try $\mathbf{r}=0$, and hope it works!
- Assuming this works for the moment, the test is

$$|\bar{y}| \Big|_{H_0} \geq \sqrt{\mathbf{I}(\mathbf{s}^2/n)^2 + \mathbf{s}^2/n} \equiv \mathbf{g}$$

Setting the Threshold

$$\mathbf{a} = P_{FA} = \Pr[|\bar{y}| > \mathbf{g} | H_0] = 1 - \Pr[-\bar{y} < \mathbf{g} < \bar{y} | H_0]$$

$$= 1 - \Pr \left[-\mathbf{g} \frac{\sqrt{n}}{\mathbf{s}} < \underbrace{\bar{y} \frac{\sqrt{n}}{\mathbf{s}}}_{\mathcal{N}(0,1)} < \mathbf{g} \frac{\sqrt{n}}{\mathbf{s}} \Big| H_0 \right]$$

$$= 2Q(\mathbf{g}\sqrt{n}/\mathbf{s})$$

$$\mathbf{g} = \frac{\mathbf{s}}{\sqrt{n}} Q^{-1}(\mathbf{a}/2)$$

Computing the Power Curve

$$P_D(\mathbf{q}) = \Pr[|\bar{y}| > \mathbf{g} | H_1] = 1 - \Pr[-\mathbf{g} < \bar{y} < \mathbf{g} | H_1]$$

$$= 1 - \Pr \left[-(\mathbf{g} + \mathbf{q}) \frac{\sqrt{n}}{\mathbf{s}} < \underbrace{(\bar{y} - \mathbf{q}) \frac{\sqrt{n}}{\mathbf{s}}}_{\mathcal{N}(0,1)} < (\mathbf{g} - \mathbf{q}) \frac{\sqrt{n}}{\mathbf{s}} \Big| H_0 \right]$$

$$= Q(-(\mathbf{g} + \mathbf{q})(\sqrt{n}/\mathbf{s})) + 1 - Q((\mathbf{g} - \mathbf{q})(\sqrt{n}/\mathbf{s}))$$

$$= Q(-Q^{-1}(\mathbf{a}/2) + d) + 1 - Q(Q^{-1}(\mathbf{a}/2) - d)$$

$$= 2 - Q(Q^{-1}(\mathbf{a}/2) + d) - Q(Q^{-1}(\mathbf{a}/2) - d)$$

where $d = \mathbf{q}\sqrt{n}/\mathbf{s}$