

## STAT 801: Mathematical Statistics

### Optimality theory for point estimates

Why bother doing the Newton Raphson steps?

Why not just use the method of moments estimates?

Answer: method of moments estimates not usually as close to right answer as MLEs.

**Rough principle:** A good estimate  $\hat{\theta}$  of  $\theta$  is usually close to  $\theta_0$  if  $\theta_0$  is the true value of  $\theta$ . Closer estimates, more often, are better estimates.

This principle must be quantified if we are to “prove” that the mle is a good estimate. In the Neyman Pearson spirit we measure average closeness.

**Definition:** The **Mean Squared Error** (MSE) of an estimator  $\hat{\theta}$  is the **function**

$$MSE(\theta) = E_{\theta}[(\hat{\theta} - \theta)^2]$$

Standard identity:

$$MSE = \text{Var}_{\theta}(\hat{\theta}) + \text{Bias}_{\hat{\theta}}^2(\theta)$$

where the bias is defined as

$$\text{Bias}_{\hat{\theta}}(\theta) = E_{\theta}(\hat{\theta}) - \theta.$$

**Primitive example:** I take a coin from my pocket and toss it 6 times. I get *HTHTTT*. The MLE of the probability of heads is

$$\hat{p} = X/n$$

where  $X$  is the number of heads. In this case I get  $\hat{p} = \frac{1}{3}$ .

Alternative estimate:  $\tilde{p} = \frac{1}{2}$ .

That is,  $\tilde{p}$  ignores data; guess coin is fair.

The MSEs of these two estimators are

$$MSE_{\text{MLE}} = \frac{p(1-p)}{6}$$

and

$$MSE_{0.5} = (p - 0.5)^2$$

If  $0.311 < p < 0.689$  then 2nd MSE is smaller than first.

For this reason I would recommend use of  $\tilde{p}$  for sample sizes this small.

Same experiment with a thumbtack: tack can land point up (U) or tipped over (O).

If I get *UOUOOO* how should I estimate  $p$  the probability of  $U$ ?

Mathematics is identical to above but is  $\tilde{p}$  is better than  $\hat{p}$ ?

Less reason to believe  $0.311 \leq p \leq 0.689$  than with a coin.