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# STATISTICS/ECONOMETRICS PREP COURSE – PROF. MASSIMO GUIDOLIN

## SECOND PART, LECTURE 4: CONFIDENCE INTERVALS

# OVERVIEW

- 1) General notion – sample (random) intervals
- 2) Coverage and confidence of intervals
- 3) Finding confidence intervals by inversion of (non-) rejection regions of tests
- 4) A few examples

# FROM POINT TO SET (INTERVAL) INFERENCES

- Point estimation of a parameter  $\theta$  boils down inference to a guess of a single value as the value of  $\theta$
- In most situations, it seems more plausible to provide a guess in the form of an interval
- Inference in a set problem is the statement that " $\theta \in \mathbb{C}$ " where  $\mathbb{C} \subset \Theta$  and  $\mathbb{C} = \mathbb{C}(\mathbf{x})$  is determined by the sample of the data
  - If  $\theta$  is real-valued, then we usually prefer the set estimate  $\mathbb{C}$  to be an interval
- Definition: An interval estimate of a real-valued parameter  $\theta$  is any pair of functions,  $L(X_1, \dots, X_n)$  and  $U(X_1, \dots, X_n)$  of a sample that satisfy  $L(\mathbf{X}) \leq U(\mathbf{X})$ ; if  $\mathbf{x}$  is observed, the inference  $L(\mathbf{x}) \leq \theta \leq U(\mathbf{x})$  is made and the **random interval**  $[L(\mathbf{x}), U(\mathbf{x})]$  is called an interval estimator
  - Although in the majority of cases we will work with finite values for

# FROM POINT TO SET (INTERVAL) INFERENCES

L and U, there is sometimes interest in one-sided interval estimates

– In the definition, using an open or closed interval,  $[L(\mathbf{x}), U(\mathbf{x})]$ , is immaterial

- What is the gain from using interval estimators? On the one hand, we have given up precision, so there must be some gain
- The answer is that we benefit from some confidence or assurance as to our assertions concerning an estimate
  - Let's stop and think: we know that the sample mean is MLE and as such unbiased, consistent, asymptotically efficient, etc. Great, but what is  $\Pr(\text{sample mean} = \mu)$ ? The answer is  $\Pr(\text{sample mean} = \mu) = 0!$ 
    - Therefore one needs to broaden the interval in order to obtain a stronger endorsement, e.g., in the case of an IID  $N(\mu, \sigma^2)$  population

$$\begin{aligned}\Pr(\mu \in [\bar{X} - a, \bar{X} + a]) &= \Pr(\bar{X} - a \leq \mu \leq \bar{X} + a) = \Pr(-a \leq \bar{X} - \mu \leq a) \\ &= \Pr\left(-\frac{a}{S/\sqrt{n}} \leq \frac{\bar{X} - \mu}{S/\sqrt{n}} \leq \frac{a}{S/\sqrt{n}}\right) = \Pr\left(-\frac{a}{S/\sqrt{n}} \leq T_{n-1} \leq \frac{a}{S/\sqrt{n}}\right)\end{aligned}$$

# COVERAGE AND CONFIDENCE OF INTERVALS

- Sets or intervals? In general one estimates confidence sets, but in finance most of the time one looks for intervals, CIs
- What confidence or assurance?
- Definition [**COVERAGE**]: For an interval estimator  $[L(\mathbf{X}), U(\mathbf{X})]$  of a parameter  $\theta$ , the coverage probability is the probability that the random interval  $[L(\mathbf{X}), U(\mathbf{X})]$  covers the true parameter,  $\theta$ , or  $\Pr(\theta \in [L(\mathbf{X}), U(\mathbf{X})])$
- Definition [**CONFIDENCE**]: For an interval estimator  $[L(\mathbf{X}), U(\mathbf{X})]$  of a parameter  $\theta$ , the confidence coefficient is the infimum of the coverage probabilities,  $\inf_{\theta} \Pr(\theta \in [L(\mathbf{X}), U(\mathbf{X})])$
- CIs are for the random sample, not for the parameter  $\theta$ : when we write statements such as  $\Pr(\theta \in [L(\mathbf{X}), U(\mathbf{X})])$ , these probability statements refer to  $\mathbf{X}$ , not  $\theta$ 
  - Re-write  $\Pr(\theta \in [L(\mathbf{X}), U(\mathbf{X})])$  as  $\Pr(L(\mathbf{X}) \leq \theta)$  together with  $\Pr(U(\mathbf{X}) \geq \theta)$

# METHODS TO FIND CONFIDENCE INTERVALS

- Since we do not know the true value of  $\theta$ , we can only guarantee a coverage probability equal to the infimum, the confidence coefficient
- Now the good news: There is a strong correspondence between hypothesis testing and interval estimation: in general, every confidence set corresponds to a test and vice versa
- So you almost, almost, do not need to separately study this...
  - Example: Let  $X_1, \dots, X_n$  be IID  $N(\mu, \sigma^2)$  and consider testing  $H_0: \mu = \mu_0$  versus  $H_1: \mu \neq \mu_0$  with  $\sigma$  known. For a fixed  $\alpha$  level, a reasonable test (in fact, the MPU) has rejection region  $\{\mathbf{x}: |\bar{x} - \mu_0| > z_{\alpha/2}\sigma/\sqrt{n}\}$  which implies that the null is not rejected for sample points with

$$\{\mathbf{x}: z_{\alpha/2}\sigma/\sqrt{n} \leq \bar{x} - \mu_0 \leq z_{\alpha/2}\sigma/\sqrt{n}\} \iff \bar{x} - z_{\alpha/2}\sigma/\sqrt{n} \leq \mu_0 \leq \bar{x} + z_{\alpha/2}\sigma/\sqrt{n}$$

Since the test has size  $\alpha$ , this means that  $\Pr(H_0 \text{ is not rejected} | \mu = \mu_0) = 1 - \alpha$ . Combining this with the characterization, we can write

$$\Pr(\bar{x} - z_{\alpha/2}\sigma/\sqrt{n} \leq \mu_0 \leq \bar{x} + z_{\alpha/2}\sigma/\sqrt{n}) = 1 - \alpha$$

# METHODS TO FIND CONFIDENCE INTERVALS

- This interval is obtained by inverting the (complement of the) rejection region of the level  $\alpha$  test
- There is no guarantee that the confidence set obtained by test inversion will be an interval
  - In most cases, however, one-sided tests give one-sided intervals, two-sided tests give two-sided intervals, strange-shaped rejection regions give strange-shaped confidence sets
- Example: As another example, we show how do you go about inverting an LRT. Suppose we want a CI for the mean,  $\lambda$ , of an exponential( $\lambda$ ) population, with PDF  $f(x; \lambda) = (1/\lambda)\exp(-x/\lambda)$ . We can obtain such an interval by inverting a size  $\alpha$  test of  $H_0: \lambda = \lambda_0$  versus  $H_1: \lambda \neq \lambda_0$ .
- If we take a random sample  $X_1, \dots, X_n$ , the LRT statistic is given by

$$LRT(\mathbf{x}) = \frac{L(\lambda_0; \mathbf{x})}{\sup_{\lambda} L(\theta; \mathbf{x})} = \frac{\frac{1}{\lambda_0^n} \exp\left(-\frac{1}{\lambda_0} \sum_{i=1}^n x_i\right)}{\sup_{\lambda} \frac{1}{\lambda^n} \exp(-n)} = \left(\frac{\bar{x}}{\lambda_0}\right)^n e^n \exp\left(1 - \frac{\bar{x}}{\lambda_0}\right)$$

- For fixed  $\lambda_0$ , the non-rejection region is given by

# METHODS TO FIND CONFIDENCE INTERVALS

$$\left\{ \mathbf{x} : \left( \frac{\bar{x}}{\lambda_0} \right)^n \exp \left( 1 - \frac{\bar{x}}{\lambda_0} \right) \geq k^* \right\} \quad (e^n \text{ absorbed into } k^*)$$

where  $k^*$  is a constant chosen to satisfy

$$\Pr\{\mathbf{x} \in \text{not reject}\} = 1 - \alpha$$

- Inverting this region gives the  $1 - \alpha$  set

$$C(\mathbf{x}) = \left\{ \lambda : \left( \frac{\bar{x}}{\lambda} \right)^n \exp \left( 1 - \frac{\bar{x}}{\lambda} \right) \geq k^* \right\} \quad (\text{generalize } \lambda_0 \text{ to } \lambda)$$

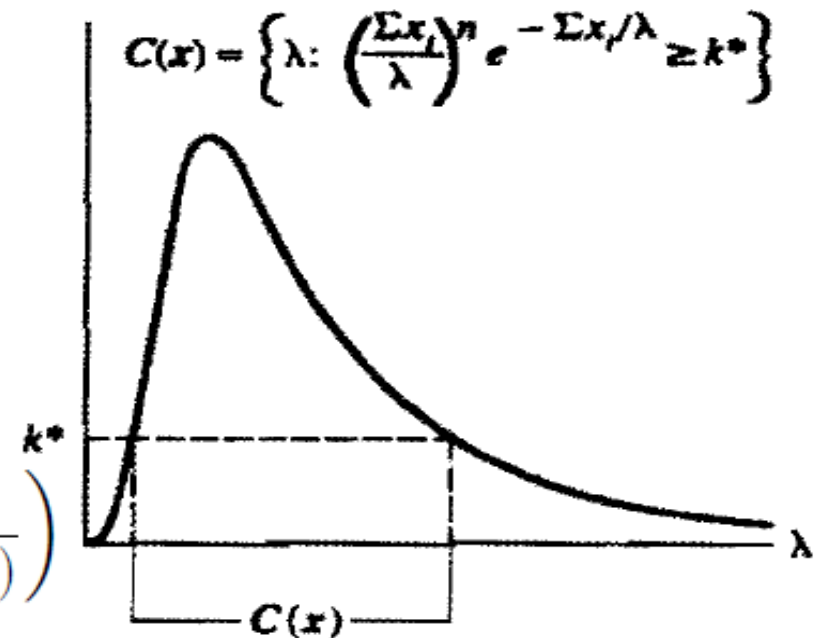
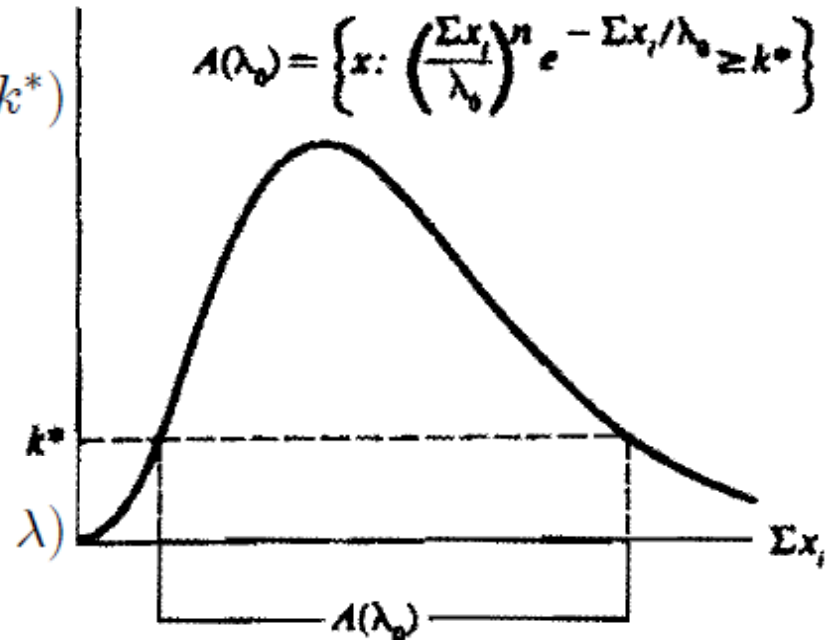
- The expression defining  $C(\mathbf{x})$  depends on  $\mathbf{x}$  only through the sample mean, so the CI can be expressed in the form

$$C(\bar{x}) = \{ \lambda : L(\bar{x}) \leq \lambda \leq U(\bar{x}) \}$$

- At this point  $L$  and  $U$  can be determined by imposing that  $\Pr\{\mathbf{x} \in \text{not reject}\} = 1 - \alpha$  and

$$\left( \frac{\bar{x}}{L(\bar{x})} \right)^n \exp \left( 1 - \frac{\bar{x}}{L(\bar{x})} \right) = \left( \frac{\bar{x}}{U(\bar{x})} \right)^n \exp \left( 1 - \frac{\bar{x}}{U(\bar{x})} \right)$$

for which the solution is numerical





# TERMINOLOGY ISSUES AND BAYESIAN CIs

- Notice that it is the CI that covers the parameter, not the opposite; in particular, our claim is not really about the probability of the parameter falling inside a CI
- This is a key remark because **the random quantity**, the sample statistic, **is the interval**, not the parameter
  - For instance, suppose a 90% CI for some parameter  $\theta$  is  $[0.24, 0.45]$
  - It is tempting to say (and many experimenters do) that "the probability is 90% that  $\theta$  is in the interval  $[0.24, 0.45]$ "
  - However, such a statement is invalid since the parameter is assumed fixed and such it cannot go anywhere, it cannot fall or rise...
  - Formally, the interval  $[0.24, 0.45]$  is one of the possible realized values of the random CI for  $\theta$  and, since the parameter does not move,  $\theta$  is in the realized interval  $[0.24, 0.45]$  with probability 0 or 1
  - That the realized interval  $[0.24, 0.45]$  has a 90% chance of coverage, we only mean that we know that 90% of the sample points of the

# USEFUL NOTIONS REVIEWED IN THIS LECTURE

- Let me give you a list to follow up to:
- Definition of random confidence interval/set
- Coverage and confidence of a confidence interval
- Terminology issue – what a CI really means