

UCLA STAT 10 Introduction to Statistical Reasoning

- **Instructor:** Ivo Dinov,
Asst. Prof. In Statistics and Neurology
 - **Teaching Assistants:** Yan Xiong and Will Anderson
UCLA Statistics
- University of California, Los Angeles, Winter 2002
<http://www.stat.ucla.edu/~dinov/>

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Slide 1

Chapter 6 – Measurement Error, Chance and Uncertainty

Slide 2

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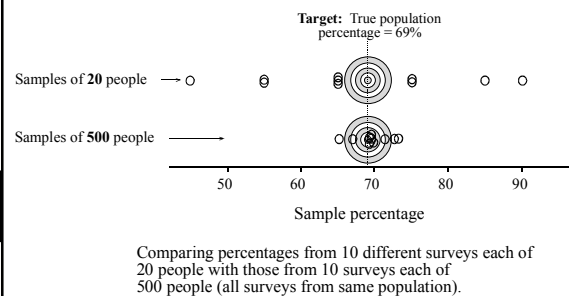
Newtonial science vs. chaotic science

- Article by Robert May, Nature, vol. 411, June 21, 2001
 - Science we encounter at schools deals with **crisp certainties** (e.g., prediction of planetary orbits, the periodic table as a descriptor of all elements, equations describing area, volume, velocity, position, etc.)
 - As soon as **uncertainty** comes in the picture it **shakes the foundation of the deterministic science**, because only **probabilistic statements** can be made in describing a phenomenon (e.g., roulette wheels, chaotic dynamic weather predictions, Geiger counter, earthquakes, Others?)
 - **What is then science all about** – describing absolutely certain events and laws alone, or describing more general phenomena in terms of their behavior and chance of occurring? Or may be both!

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Variation in sample percentages



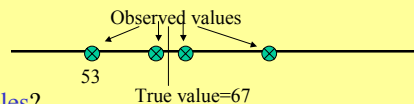
From Chance Encounters by C.J. Wild and G.A.F. Seber, © John Wiley & Sons, 2000.

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Measurement Error

- No matter how carefully a measurement of a single unit is made it often comes out a bit different. Do repeated measurements to find out by how much different each observation is!
- The SD of a series of repeated measurements estimates the **likely size of the chance error** in a single measurement of the process being observed.



- **Examples?**

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Random or chance error ...

- **Random or chance error** is the difference between the **sample-value** and the **true population-value** (e.g., 53 vs. 67, in the above example).

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The Subject of Statistics

Statistics is concerned with the process of finding out about the world and how it operates -

- in the face of **variation** and **uncertainty**
- by **collecting** and then **making sense** (interpreting, summarizing) of data.

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The investigative process

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Questions

- What are two ways in which random observations arise and give examples. (random sampling from finite population – randomized scientific experiment; random process producing data, observational data, surveys.)
- What is a **parameter**? Give two examples of parameters. (characteristic of the data – mean, 1st quartile, std.dev.)
- What is an **estimate**? How would you estimate the parameters you described in the previous question?
- What is the distinction between an **estimate** (p^{\wedge} value calculated from obs'd data to approx. a parameter) and an **estimator** (P^{\wedge} abstraction of the properties of the random process and the sample that produced the estimate) ? Why is this distinction necessary? (effects of sampling variation in P^{\wedge})

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Review

- Let $\{X_{1,1}, X_{1,2}, X_{1,3}, \dots, X_{1,N}\}$
 $\{X_{2,1}, X_{2,2}, X_{2,3}, \dots, X_{2,N}\}$
 \dots
 $\{X_{K,1}, X_{K,2}, X_{K,3}, \dots, X_{K,N}\}$

K samples of size N. Data comes from a distr'n with μ, σ , but we're interested in mean/std-dev of sample average

- As the number of samples and the number of observations within each sample increase we get a better estimate of the true population parameter (say the mean). Scottish soldiers chest measurements example ...

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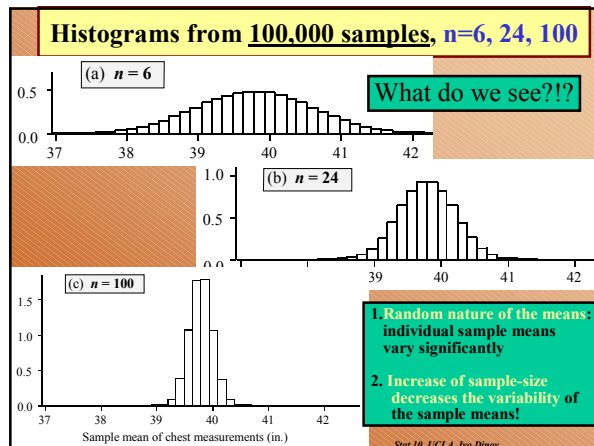
The sample mean has a sampling distribution

Sampling batches of 6 Scottish soldiers and taking chest measurements. Population $\mu = 39.8$ in, and $\sigma = 2.05$ in.

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Twelve samples of size 24

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Mean and SD of the sampling distribution

$E(\text{sample mean}) = \text{Population mean}$

$$SD(\text{sample mean}) = \frac{\text{Population SD}}{\sqrt{\text{Sample size}}}$$

$$E(\bar{X}) = E(X) = \mu, \quad SD(\bar{X}) = \frac{SD(X)}{\sqrt{n}} = \frac{\sigma}{\sqrt{n}}$$

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Review

- We use both \bar{x} and \bar{X} to refer to a sample mean. For **what purposes** do we use the *former* and for what purposes do we use the *latter*? (sample mean *estimate* vs. *estimator*)
- What is meant by “the **sampling distribution** of \bar{X} ”?
(*sampling variation* – the observed variability in the *process of taking random samples*;
sampling distribution – the real probability distribution of the random sampling process)
- How is the **population mean of the sample average** \bar{X} related to the **population mean of individual observations**? ($E(\bar{X}) = \text{Population mean}$)

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Bias and Precision

- The **bias** in an estimator is the **distance between the center of the sampling distribution of the estimator and the true value of the parameter** being estimated. In math terms, **bias** = $E(\hat{\theta}) - \theta$, where theta, $\hat{\theta}$, is the estimator, as a RV, of the true (unknown) parameter θ .
- Example, Why is the **sample mean** an **unbiased** estimate for the **population mean**? How about $\frac{3}{4}$ of the sample mean?

$$E(\hat{\theta}) - \mu = E\left(\frac{1}{n} \sum_{k=1}^n X_k\right) - \mu = 0$$

$$E(\hat{\theta}) - \mu = E\left(\frac{3}{4} \frac{1}{n} \sum_{k=1}^n X_k\right) - \mu = \frac{3}{4} \mu - \mu = \frac{\mu}{4} \neq 0, \text{ in general.}$$

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Bias and Precision

- $X = \{1, -1, 3\}$, sample_mean = 1, Estimator1 = $\frac{3}{2}$
xSampleMean = 1.5

$$E(\hat{\theta}) - \mu = E\left(\frac{1}{n} \sum_{k=1}^n X_k\right) - \mu = 0$$

$$E(\hat{\theta}) - \mu = E\left(\frac{3}{4} \frac{1}{n} \sum_{k=1}^n X_k\right) - \mu = \frac{3}{4} \mu - \mu = \frac{\mu}{4} \neq 0, \text{ in general.}$$

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Bias and Precision

- The **precision** of an estimator is a **measure of how variable is the estimator in repeated sampling**.

(a) No bias, high precision (b) No bias, low precision

(c) Biased, high precision (d) Biased, low precision

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Standard error of an estimate

The *standard error* of any estimate $\hat{\theta}$ [denoted $se(\hat{\theta})$]

- estimates the variability of $\hat{\theta}$ values in repeated sampling and
- is a measure of the *precision* of $\hat{\theta}$. Example: \bar{X} , as an estimator of the population mean, μ .

$$SE(\bar{X}) = \frac{\sigma}{\sqrt{n}}, \text{ where } \bar{X} = \frac{1}{n} \sum_{k=1}^n X_k, \text{ and}$$

σ is the standard deviation of $\{X_k\}$, $1 \leq k \leq n$.

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Review

- What is meant by the terms **parameter** and **estimate**.
- Is an estimator a RV?
- What is **statistical inference**? (process of making conclusions or making useful statements about unknown distribution parameters based on observed data.)
- What are **bias** and **precision**?
- What is meant when an estimate of an unknown parameter is described as **unbiased**?

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