Populations, Samples, & Data Summary in Nat. Resource Mgt.

ESRM 304

Sampling in Natural Resources Management

- I. Basic Concepts
- II. Tools of the Trade
- III. A Most Important Distribution

- A. Why sample?
- B. Populations, Parameters, Estimates
- C. Variables continuous, discrete
- D. Bias, Accuracy, Precision
- E. Distribution functions

- A. Why sample?
 - 1. Partial knowledge is a normal state
 - 2. Complete enumeration is impossible
 - 3. Complete enumeration is too expensive
 - 4. Results are needed in a timely manner

- B. Populations, Parameters, Estimates
 - 1. Population: An aggregate of unit values
 - 2. Parameter: A constant used to characterize a particular population
 - 3. Estimate: A value calculated from a sample in a way that makes it a 'good' approximation to a parameter

Statistic: A value calculated from a sample

- C. Variables continuous & discrete
 - Continuous: A variable that can be measured using a numerical scale that can be subdivided, if desired, into an infinite number of smaller values
 - 2. Discrete: Two (2) types:
 - a) Attributes: binomial –or– multinomial
 - b) Counts

- D. Bias, Accuracy, Precision
 - 1. Bias:- Systematic distortion
 - 2. Accuracy:- Nearness to true (or population) value
 - 3. Precision:- clustering of unit values to their own mean



E. Distribution functions Show for a sample (or population) the relative frequency with which different values occur



Another way to look at Bias, Accuracy, Precision μ Reference value Probability Accuracy density Value Precision

- A. Subscripts, Summations, Brackets
- B. Mean, Variance, Standard Deviation
- C. Standard Error of the estimate
- D. Coefficient of Variation
- E. Covariance, Correlation

A. <u>Subscripts</u>, Summations, Brackets
A subscript can refer to a unit in a sample, *e.g.*,
x₁ is value on 1st unit, x₂ is value on of 2nd, etc.,
... it can refer to different populations of values, *e.g.*,
x₁ can refer to the value tree height, while x₂ can refer to the value tree diameter,

... there can be more than one subscript, *e.g.*, x_{ij} may refer to the jth individual of the ith species of tree, where j = 1, ..., 50; i = DF, WH, RC

A. Subscripts, <u>Summations</u>, Brackets To indicate that several (say 6) values of a variable, x, are to be added together, we could write $(x_1 + x_2 + x_3 + x_4 + x_5 + x_6)$ or shorter $(x_1 + x_2 + \dots + x_6)$ shorter still

$$\sum_{i=1}^{6} x_i$$
 or even $\sum_i x_i$ or just $\sum x_i$

A. Subscripts, Summations, <u>Brackets</u> Order of operations still apply using "sigma" notation, *e.g.*,

$$\sum_{i=1}^{3} x_i y_i = x_1 y_1 + x_2 y_2 + x_3 y_3$$

$$\left(\sum_{i=1}^{3} x_i\right) \left(\sum_{i=1}^{3} y_i\right) = (x_1 + x_2 + x_3)(y_1 + y_2 + y_3)$$

$$\left(\sum_{i=1}^{3} x_i^2\right) \neq \left(\sum_{i=1}^{3} x_i\right)^2 \text{ i.e., } (x_1^2 + x_2^2 + x_3^2) \neq (x_1 + x_2 + x_3)^2$$

B. Mean, Variance, Standard Deviation Mean: $\overline{x} = \frac{1}{n} \left(\sum_{i=1}^{n} x_i \right) = \frac{1}{n} \sum_{i=1}^{n} x_i$ Variance: $s_x^2 = \frac{\sum_{i=1}^{n} (x_i - \overline{x})^2}{(n-1)} = \frac{\sum_{i=1}^{n} x_i^2 - \frac{1}{n} \left(\sum_{i=1}^{n} x_i \right)^2}{(n-1)}$

Standard Deviation: $S_x = \sqrt{S_x^2}$

B. Mean, Variance, Standard Deviation - Example
 Let's say we have measurements on 3 units sampled from a large population. Values are 7, 8, and 12 ft.

$$\overline{x} = \frac{1}{n} \sum_{i=1}^{n} x_i = \frac{1}{3} (7 + 8 + 12) = 9 ft$$
$$s_x^2 = \frac{(7^2 + 8^2 + 12^2) - \frac{1}{3} (7 + 8 + 12)^2}{2} = 7 ft^2$$

 $S_x = \sqrt{S_x^2} = \sqrt{7 ft^2} = 2.64 ft$

C. Standard Error of an estimate

- The most frequently desired estimate is for the mean of a population
- ✓ We need to be able to state how reliable our estimate is
- ✓ Standard error is key for stating our reliability
- Standard error quantifies the dispersion between an estimate derived from different samples taken from the same population of values
- ✓ Standard deviation of the observations is the square root of their variance, standard error (of an estimate) is the square-root of the variance of the estimate

Standard Error of an estimate - Example С. Let's say we have a population of (N = 15) tree heights: 7, 10, 8, 12, 2, 6, 5, 9, 3, 7, 4, 8, 9, 11, 5 from which we take 4 units (n = 4) five separate times ... pick 1 (units 10, 8, 3, 11): 7, 9, 8, 4; $\bar{x} = 7; s = 2.16$ pick 2 (units 5, 3, 6, 4) : 2, 8, 6, 12; $\overline{x} = 7$; s = 4.16pick 3 (units 8, 11, 3, 13): 9, 4, 8, 9; $\overline{x} = 7.5$; s = 2.38 $\overline{x} = 5; s = 4.08$ pick 4 (units 9, 14, 11, 5): 3, 11, 4, 2; $\overline{x} = 6.75; s = 3.40$ pick 5 (units 5, 3, 2, 10) : 2, 8, 10, 7; ... there are 1,365 possible unique samples of size 4 !!!

C. Standard Error of an estimate - Example (cont' d)
 If we used Simple Random Sampling (SRS), there is a very direct way to calculate standard error of the estimated (sample) mean

In words: standard deviation divided by the square-root of the sample size

In formula: $s_{\bar{x}} = \frac{S_x}{\sqrt{n}}$

pick 1: 1.08; 2: 2.08; 3: 1.19; 4: 2.04; 5: 1.70 Population mean = 7.07; std.dev = 2.91; std.err = 1.457

D. Coefficient of Variation

- Puts variability on a relative scale so we can compare the dispersions of values measured in different units (say feet and meters) or the dispersion of different populations (say heights and weights)
- \checkmark Ratio of standard deviation to the mean

D. Coefficient of Variation - Example Using the previous tree height population ... pick 1: $\overline{x} = 7$; s = 2.16

$$C = \frac{s}{\overline{x}} = \frac{2.16}{7} = 0.308$$
 or, ~ 31 %

If inches had been used, $\overline{x} = 84$; s = 25.92 $C = \frac{s}{\overline{x}} = \frac{25.92}{84} = 0.308$

E. <u>Covariance</u>, Correlation

- ✓ In some situations, we'd like to know if two variables (call one x, the other y) are associated with each other
- \checkmark If the association is direct, covariance is positive
- ✓ If indirect, covariance is negative
- \checkmark If not associated, covariance is nearly zero

$$s_{xy} = \frac{\sum_{i=1}^{n} (x_i - \overline{x}) (y_i - \overline{y})}{(n-1)} = \frac{\sum_{i=1}^{n} x_i y_i - \frac{1}{n} \left(\sum_{i=1}^{n} x_i\right) \left(\sum_{i=1}^{n} y_i\right)}{(n-1)}$$

Covariance - Example E.

> We have a sample of units from a population on which we measured values of two variables

> > 14.4

6 - 1

E. Covariance, Correlation

- As with variance, the magnitude of the covariance can be related to magnitude of the unit values
- A measure of the degree of association that is unaffected by size of unit values (like coefficient of variation) is the correlation coefficient
 - Correlation coefficient varies between -1 and +1
 - Closer it is to 1 (either sign), the stronger the association it is

E. Correlation - Example

$$r = \frac{\text{covariance of x and y}}{\sqrt{(\text{variance of x})(\text{variance of y})}} = \frac{s_{xy}}{\sqrt{(s_x^2)(s_y^2)}}$$

$$r = \frac{s_{xy}}{\sqrt{(s_x^2)(s_y^2)}} = \frac{-14.4}{\sqrt{(12.0)(18.4)}} = -0.969$$

The Normal Distribution

Greek symbols denote parameters: Mean: μ Variance: σ^2

English (latinbased) letters denote statistics:

 S^2

 \mathcal{X}



Properties of the Normal Distribution

- The distribution is bell-shaped; symmetrical about mean
- The mean locates the center of the distribution.
- The standard deviation is the distance between the mean and the inflection point of the distribution function.
- The distribution covers the entire real number line, from $-\infty$ to $+\infty$
- It has two parameters: the mean, μ and variance, σ^2

A couple of Normal Distributions



Why all the fuss about the Normal?

It has a variety of uses:

- Many populations found in nature are distributed approximately this way
- Used to calculate the chances a value within a certain range will occur
- Describing experimental error (calculating confidence)
- The distribution of sample means is approximately Normal (Central Limit Theorem)

Why all the fuss about the Normal?

Used to calculate the chances a particular value will be observed within a population (or a range of values)

- Any random variable X following a Normal distribution with mean = μ and variance = σ^2 can be 'mapped' onto the so-called *Standard* Normal (or "Z" distribution, which has a mean of zero and a variance of one) by the following equation:

$$Z = \frac{X - \mu}{\sigma}$$

The Central Limit Theorem:

If the mean, \overline{X} of a random sample $X_1, X_2, X_3, ..., X_n$ of size n arising from ANY distribution with a finite mean and variance is transformed into W, using the following equation:

$$W = \frac{\overline{X} - \mu}{\sigma_{\overline{X}}}$$
, where $\sigma_{\overline{X}} = \sqrt{\sigma^2/n}$

the distribution of W will *approach* that of a standard Normal deviate with mean = 0, and variance = 1 in the "limit," i.e., as sample size $n \rightarrow \infty$.

The Normal distribution does have its limits...

- Application of the normal dist' n assumes σ is known
 - Using it with unknown s.d. will overstate confidence & reliability, especially when we also have a small sample (n < ?)
- When we do not know population standard deviation (or variance), use *Student's t* distribution instead
 - The "t" distribution should be used *especially* when we also have a small sample
- Like the normal, "t" is symmetrical, spans $-\infty$ to $+\infty$
- Unlike the normal, a single parameter defines it, *V*, i.e., the so-called *degrees of freedom (or df)*

The Central Limit Theorem (unknown σ) If the mean, \overline{X} of a random sample $X_1, X_2, X_3, ..., X_n$ of size *n* (where *n* is small) from a population distributed as a Normal is transformed into W, using the following equation:

$$W = \frac{\overline{X} - \mu}{S_{\overline{X}}}$$
, where $S_{\overline{X}} = \sqrt{S^2/n}$

the distribution of W follows the "Student's t" distribution. If the sample is large enough, W will still map onto the standard Normal (or "Z" distribution) even with unknown variance and unknown population dist'n

Things to Remember- Sampling in Nat. Resources Management

- I. Basic Concepts
 - ✓ Populations have parameters
 - ✓ Samples have statistics (to estimate parameters)
- II. Tools of the Trade
 - ✓ Standard deviation is the square-root of variance
 - ✓ Standard deviation (sd) and Standard Error (se) both quantify dispersion
 - SD for dispersion of sample values
 - SE for dispersion of sample mean values

Things to Remember- Sampling in Nat. Resources Management

III. A Most Important Distribution Function

- The normal distribution has nice properties for describing a population of values measured on a continuous scale (number line)
- The "Normal" does not do everything for us; we need to use the "t" distribution when pop'n variance is unknown and especially when we have small samples