

- e. Probability Distribution: Add next to coins frequency chart a P(x) with 1/8, 3/8, 3/8, 1/8 values
- f. Probability Function: Obey two properties of prob. (0 ≤ P(A) ≤ 1, ∑ (all outcomes) P(A) = 1.
- g. Parameter: Unknown # describing population
- h. Statistic: # computed from sample data

	Sample	Population
Mean	$\bar{x}$	$\mu$ - mu
Variance	$s^2$	$\sigma^2$
Standard deviation	$s$	$\sigma$ - sigma

i. Base:  $\bar{x} = \sum x/n, s^2 = \frac{\sum (x - \bar{x})^2}{(n - 1)}$

	Frequency Dist.	Probability Distribution
Mean	$\bar{x} = \sum xf / \sum f$	$\mu = \sum [xP(x)]$
Var	$s^2 = \frac{\sum (x - \bar{x})^2 f}{(\sum f - 1)}$	$\sigma^2 = \sum [(x - \mu)^2 P(x)]$
Std Dev	$s = \sqrt{s^2}$	$\sigma = \sqrt{\sigma^2}$

- j. Probability acting as an  $f / \sum f$ . Lose the -1

**9. Sampling Distribution**

- a. By law of large #s, as  $n \rightarrow$  population,  $\bar{x} \rightarrow \mu$
- b. Given  $\bar{x}$  as mean of SRS of size  $n$ , from pop with  $\mu$  and  $\sigma$ . Mean of sampling distribution of  $\bar{x}$  is  $\mu$  and standard deviation is  $\sigma / \sqrt{n}$
- c. If individual observations have normal distribution  $N(\mu, \sigma)$  - then  $\bar{x}$  of  $n$  has  $N(\mu, \sigma / \sqrt{n})$
- d. Central Limit Theorem: Given SRS of  $b$  from a population with  $\mu$  and  $\sigma$ . When  $n$  is large, the sample mean  $\bar{x}$  is approx normal.

**10. Binomial Distribution**

- a. Binomial Experiment. Emphasize Bi - two possible outcomes (success, failure).  $n$  repeated identical trials that have complementary  $P(\text{success}) + P(\text{failure}) = 1$ . binomial is count of successful trials where  $0 \leq k \leq n$
- b.  $p$ : probability of success of each observation
- c. Binomial Coefficient:  $nCk = n! / (n - k)! k!$
- d. Binomial Prob:  $P(x = k) = \binom{n}{k} p^k (1 - p)^{n - k}$
- e. Binomial  $\mu = np$
- f. Binomial  $\sigma = \sqrt{np(1 - p)}$

**11. Confidence Intervals**

- a. Statistical Inference: methods for inferring data about population from a sample
- b. If  $\bar{x}$  is unbiased, use to estimate  $\mu$
- c. Confidence Interval: Estimate  $\pm$  error margin
- d. Confidence Level C: probability interval captures true parameter value in repeated samples
- e. Given SRS of  $n$  & normal population, C confidence interval for  $\mu$  is:  $\bar{x} \pm z^* \sigma / \sqrt{n}$
- f. Sample size for desired margin of error - set  $\pm$  value above & solve for  $n$ .

**12. Tests of Significance**

- a. Assess evidence supporting a claim about popu.
- b. Idea - outcome that would rarely happen if claim were true evidences claim is not true
- c.  $H_0$  - Null hypothesis: test designed to assess evidence against  $H_0$ . Usually statement of no effect
- d.  $H_a$  - alternative hypothesis about population parameter to null
- e. Two sided:  $H_0: \mu = 0, H_a: \mu \neq 0$
- f. P-value: probability, assuming  $H_0$  is true, that test statistic would be as or more extreme (smaller P-value is  $\rightarrow$  evidence against  $H_0$ )
- g.  $z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$
- h. Significance level  $\alpha$ : If  $\alpha = .05$ , then happens no more than 5% of time. "Results were significant ( $P < .01$ )"
- i. Level  $\alpha$  2-sided test rejects  $H_0: \mu = \mu_0$  when  $u_0$  falls outside a level  $1 - \alpha$  confidence int.
- j. Complicating factors: not complete SRS from population, multistage & many factor designs, outliers, non-normal distribution,  $\sigma$  unknown.
- k. Under coverage and nonresponse often more serious than the random sampling error accounted for by confidence interval
- l. Type I error: reject  $H_0$  when it's true -  $\alpha$  gives probability of this error
- m. Type II error: accept  $H_0$  when  $H_a$  is true
- n. Power is  $1 -$  probability of Type II error