MATH 10

INTRODUCTORY STATISTICS

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Tentative Plans For Midterm Exam

- Week 5 26 April (Thursday), 1 hour 45 minutes.
- In class, here in this room.
- Tentatively looking at 6 questions, 15 minutes per question.
- Will confirm details next week.
- This is a fast paced course. If you feel that you're lagging behind, I can review key points and work through sample problems during office hours (or by appointment).

Week 3

- Chapter 5 Probability
- Chapter 7 Normal Distribution

• Chapter 9 – Sampling Distributions

← today's lecture

Sampling distributions of the mean and p. Difference between means. Central Limit Theorem.

Chapter 8 – Advanced Graphs

← today's lecture

Sample exam problem with Bayes Theorem

Police officers in a town are stopping drivers at random and subjecting them to a breathalyzer test that detects alcohol intoxication.

During the operation, there were 5000 drivers on the road. One of those drivers were chosen at random, say Tommy, and he tested positive (>_< !!!).

Sample exam problem with Bayes Theorem

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1. The police wants to deport Tommy because they think the probability that he is driving drunk is 99%. Is their claim true or false based on the information so far? Explain. (2 pts)

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Now you are told that out of the 5000 drivers on the road, 5 of them are drunk, while the rest are sober.

2. If all 4995 sober drivers are given the breathalyzer test, approximately how many would get a positive result? (1 pt)

3. The police was picked a driver at random from the 5000. What is the probability that he or she is sober? You do not have to simplify your answer. (1 pt)

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Now you are told that out of the 5000 drivers on the road, 5 of them are drunk, while the rest are sober.

4. Using Bayes Theorem, write down a numerical fraction for the probability that a randomly selected driver is not drunk, given that he or she got a positive result. (4 pts)

- Take any interval [*a*, *b*].
- The area under the curve, within this interval, is the probability that a normally distributed variable has value in [*a*, *b*].

Normal distribution mean = 100 standard deviation = 20

Blue area = 0.68 or 68%



- Take any interval [*a*, *b*].
- The area under the curve, within this interval, is the probability that a normally distributed variable has value in [*a*, *b*].

• Alternatively, for an approximately normal set of data, the normal curve tells us the proportion of data with values within [a,b].

Normal distribution mean = 100 standard deviation = 20

Blue area = 0.68 or 68%



- The "68 95 99.7" heuristic.
- You should know this for the exam.

• ...or at remember enough to look it up a z-value table.



The 68-95-99.7 rule in practice for an approximately normal histogram. Image from Wikipedia.

Example exam question:

• The proportion of scores for 1,000 students in a class are well approximated by a normal distribution, with mean 50 and standard deviation 10.

• Approximately how many students scored 70 and above? (2 pts)

• Show your work. Illustrations are allowed.

Chapter 7, Section 6 – Standard Normal

• Alternative way to solve this question.

- Let X be a normal distributed (random) variable with mean μ and variance σ^2 .
- Apply the linear transformation: $Z = \frac{X \mu}{\sigma}$.
- Now Z is standard normal.
- \rightarrow normally distributed, mean 0, variance 1

Chapter 7, Section 6 – Standard Normal

- Apply the linear transformation: $Z = \frac{X \mu}{\sigma}$.
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- The proportion of scores for 1,000 students in a class are well approximated by a normal distribution, with mean 50 and standard deviation 10.
- Approximately how many students scored 70 and above? (2 pts)

• Will be used a lot for hypothesis testing later in the course.

Let's try a tougher sample exam question:

• The weights for 500 bars of gold are well approximated by a normal distribution, with mean 100 grams and standard deviation 20 grams.

• Approximately how bars of gold weigh between 100 grams and 130 grams inclusive? (2 pts)

- The weights for 500 bars of gold are well approximated by a normal distribution, with mean 100 grams and standard deviation 20 grams.
- Approximately how bars of gold weigh between 100 grams and 130 grams inclusive? (2 pts)

- Two ways to think about this...
- 1. Area under curve as proportion of those 500 bars of gold.
- 2. Area under curve as the probability of getting a bar of gold within certain weights.

Break time!! \o/

• Break starts after I hand out the exercise.

• Circle is a timer that becomes blue. O_o (please ignore if it glitches) 12 minutes



 \rightarrow

Chapter 9, Section 2 - Introduction

- Inferential statistics.
- Have population, and a variable of interest, X.
- Take a simple random sample, calculate estimators for certain aspects of the population distribution of X.

- E.g. sample mean = estimator for population mean.
- E.g. estimator of the sample variance.

• We will now quantify how "good" these estimators are.

• The sample mean, of a simple random sample A with size *n*, is a random variable.

• If you collect another simple random sample B with size n, it is likely to have a different sample mean.

• If \overline{X}_n is a random variable that represents the mean of a sample of size n, then \overline{X}_n has a distribution.

• The distribution of \overline{X}_n is the sampling distribution of the mean (of a sample of size n).

• This distribution has mean $\mu_M = \mu$, where μ is the population mean.

• This distribution has variance $\sigma_M^2 = \frac{\sigma^2}{n}$, where σ^2 is the population variance.

• Sampling Distribution of the Mean has,

$$\mu_M = \mu \qquad \sigma_M^2 = \frac{\sigma^2}{n}$$

• Standard error, $\sigma_M = \frac{\sigma}{\sqrt{n}}$.

• Sampling Distribution of the Mean has,

$$\mu_M = \mu$$
 $\sigma_M^2 = rac{\sigma^2}{n}$ error, $\sigma_M = rac{\sigma}{\sqrt{n}}$.

• Standard error,
$$\sigma_M = \frac{\sigma}{\sqrt{n}}$$
.

• Central Limit Theorem!!! ふ(ゔ)タ

If the population has finite mean μ , and finite non-zero variance σ^2 , then the sampling distribution of the mean becomes better approximated by a normal distribution N(μ , $\frac{\sigma^2}{n}$), as sample size n increases.

• Central limit theorem works for **any** distribution with finite mean and finite non-zero variance.



Chapter 9, Section 7 – Difference Between Means

• Finally, we can use statistics to compare two populations.

- Suppose you have two simple random samples with size n_1 and n_2 .
- Samples from population 1 and 2 respectively.

- Calculate their sample means M_1 and M_2 .
- The difference has a sampling distribution with mean

$$\mu_{M_1-M_2} = \mu_1 - \mu_2.$$

Chapter 9, Section 7 – Difference Between Means

- The difference has a sampling distribution with mean $\mu_{M_1-M_2} = \mu_1 \mu_2$.
- And variance $\sigma_{M_1-M_2}^2 = \sigma_{M_1}^2 + \sigma_{M_1}^2$.

•
$$\sigma_{M_i}^2 = \frac{\sigma^2}{n_i}$$
, which is variance of the sampling distribution of M_i .

• Since the sample means are independent (as random variables), the variance sum law was used to derive the variance.

•
$$\sigma_{M_1-M_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

Chapter 9, Section 7 – Difference Between Means

• The difference has a sampling distribution with mean $\mu_{M_1-M_2} = \mu_1 - \mu_2$.

• And variance
$$\sigma_{M_1-M_2}^2 = \sigma_{M_1}^2 + \sigma_{M_1}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$
.

• Standard error
$$\sigma_{M_1-M_2} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

• This becomes much easier if the sample sizes and population variances are equal.

Public Service Announcement

- We are skipping "Chapter 9, Section 8, Sampling Distribution of r".
- This chapter is about the sampling distribution of the correlation coefficient.
- Not usually taught at Math 10 level.
- So we're nuking it from orbit (it's the only way to be sure).