Name _		
Per	Date	

9.1A	A basketball player claims to make 80% basketball before and you don't believe	of the free throws he attempts. You have seen him play him. So you ask him to shoot some free throws. Suppose he		
	shoots 50 free throws and makes 32 of them. Is his claim true but he just had a "bad streak" or is his claim false?			
	A) What is his sample proportion (p)?			
	B) what can we conclude about t	the player's claim based on the sample data?		
	When you are an 80% free th Instead we average all perform So how do we determine th We can perform a simulation	row shooter you don't always make 80% of your free throws. mances of shooting free throws to get the 80%. e likelihood (probability) that he had a "bad streak"? to find out. We used Eathom software to simulate 400 sets of		
	50 shots assuming that the pl results is below. Each dot is	ayer really is an 80% free throw shooter. The dotplot of the the proportion of shots made for each group of 50 attempts.		
	C) How many of the 400 sets of shots we			
		sample proportion () or lower?		
	In 400 sets of 50 shots, there were only 3 times when our shooter made as few as the observed	You can say how strong the evidence against the player's claim is by giving the that he would make as few as 32 out of 50 free throws if he really makes 80% in the long run.		
		D) What is the probability that he made only 32 shots		
		out of 50 tries JUST BY CHANCE ("bad streak")? Based on the simulation, our estimate of this probability is		
	0.5 0.6 0.7 0.8 0.9 p	The observed statistic, $p = 0.64$, is		
	$\hat{p} = 0.64$	that it gives convincing evidence that the player's claim is		
	Be sure that you understand why this evidence is convincing. There are two possible explanations of the fact that our virtual player made only $p = 0.64$ of his free throws: 1. The player's claim is correct ($p = 0.8$), and by bad luck, a very unlikely outcome occurred.			
	2. The population proportion is actu	ually less than 0.8, so the sample result is not an unlikely		
	outcome.			
9.1A	Significance test A formal procedure for comparing	with a whose		
	truth we want to assess. We express the	e results of a significance test in terms of a probability that		
	Significance tests	im agree.		
	Deal with claims about a			
	 "If we took many random sample 	es and the claim were true, we would get a result like this%		
	 of the time" BASIC IDEA: An outcome that w 	yould rarely happen if a claim were true is good evidence that		
	the claim is			

9.1A	Stating a hypotheses
	 <u>Null hypothesis</u> (H_o) a statement of(What the person claims to be true)
	Will always be in the form H_0 : $p = __$ or H_0 : $\mu = __$
	 <u>Alternative hypothesis</u> (H_a) the claim that we hope or instead of the
	null hypothesis.
	 <u>One sided</u> alternative hypothesis will claim that the actual parameter is
	than the null hypothesis.
	Will always be in the form H_a : p > or H_a : μ > or H_a : p < or H_a : μ <
	 <u>Two sided</u> alternative hypothesis will claim that the actual parameter is the
	parameter stated in the null hypothesis.
	Will always be in the form H_a : $p \neq ___$ or H_a : $\mu \neq ___$
	****Note the number in the null and alternative hypothesis will always be the same!
	Sample: Mike is an avid golfer who would like to improve his play. A friend suggests getting new clubs and lets Mike try out his 7-iron. Based on years of experience, Mike has established that the mean distance that balls travel when hit with his old 7-iron is $\mu = 175$ yards with a standard deviation of $\sigma = 15$ yards. He is boning that this new club will make his shots with a 7-iron more consistent (less
	spread), so he goes to the driving range and hits 50 shots with the new 7-iron.
	Problem:
	(a) State the appropriate null hypotheses
	(b) State the appropriate alternative hypotheses
9.14	Sample: VitaBland, a new brand of vitaming, claims in a radio spot that people who take VitaBland
3.1A	vitamins wake refreshed and ready for the new day within five minutes of their alarm sounding. VitaBlend's internal research department conducts a test to support the claim heard on the radio. a.) Null hypothesis: H ₀ :
	b.) Alternative hypothesis: H_a :
0.1.0	Compley The appendix inclusion is not placed with the first competer even econes of the two first period
9.1A	US history classes. The mean test score of both classes was 72. He believes that the student's final exam score average will increase because the two teachers will teach the second semester by combining the classes and using a "team teaching" approach. a.) Null hypothesis: H ₀ :
	h) Altornativa hypothasia: H :
	b.) Alternative hypothesis. Ha.
9.1A	Sample: Heading into the election, Jack needs 50% of the vote (or higher) to win the election. Jack is feeling pretty confident that he will win the election. Is his confidence warranted? a.) Null hypothesis: H ₀ :
	b.) Alternative hypothesis: H_a :

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9.1A	Notes about hypotheses:		
	The alternative hypothesis should express the hopes or suspicions we have we		
	see the data. It is cheating to look at the data first and then frame the hypotheses to fit		
	what the data show.		
	 Hypotheses always refer to a, not to a 		
	If unsure, use a alternative hypothesis!		
9.1A	Interpreting P-values		
	The probability that measures the against a null hypothesis is called a		
	The the p-value, the stronger the evidence against the null hypothesis		
	provided by the data.		
0.1.0	I ne p-value is the conditional probability Significance Level		
9.1A	<u>Significance Level</u>		
	To determine what P-value is considered small () or large () we		
	Compare it to a (d).		
	Significance level requires evidence against Ho to be that is would		
	happen than% of the time by chance when H_o is true.		
	 When our P-value is our chosen α, we say that the result is 		
	statistically significant.		
	 Significant doesn't mean important, it means not likely to 		
	 If you are going to draw a conclusion based on statistical significance, the significance level 		
	should be determined the data are produced.		
	Significance levels should range from Courts will accept 0.05 or lower.		
	When in doubt or if not explicitly stated in a problem, use		
9.1A	Statistical significance		
	The final step in performing a significance test is to draw a conclusion about the competing claims you		
	were testing. We will make one of based on the strength of the evidence:		
	•		
	• — Does not guarantee that H_0 is just that there is not		
	enough evidence to reject.		
	***DO NOT ACCEPT Ho. you will lose credit on AP exam!		
9.1A	In the free-throw shooter example , the estimated p-value of 0.0075 is against		
•••••	the null hypotheses H_0 : $p = 0.80$. For that reason, we would in favor of the		
	alternative hypotheses H_a : $p < 0.80$. It appears that the virtual player makes fewer than 80% of his free		
	throws.		
9.1A	Sample: When Mike was testing a new 7-iron, the hypotheses were		
	$H_{g}: \sigma = 15$		
	H_a : $\sigma < 15$		
	where σ = the true standard deviation of the distances Mike hits golf balls using the new 7-iron. Based		
	on 50 shots with the new 7-iron, the standard deviation was $s_x = 10.9$ yards.		
	Problem: A significance test using the sample data produced a <i>P</i> -value of 0.002.		
	(a) Interpret the <i>P</i> -value in this context.		
	(b) Our significance level is 0.01, what should our conclusion be?		

9.1A	Sample: For his second semester project in AP Statistics, Zenon decided to investigate if students at his school prefer name-brand potato chips to generic potato chips. He randomly selected 50 students and had each student try both types of chips, in random order. Overall, 34 of the 50 students preferred the name-brand chips. Zenon performed a significance test using the hypotheses: $H_0: p = 0.5$ $H_a: p > 0.5$ where p = the true proportion of students at his school that prefer name-brand chips. The resulting <i>P</i> -value was 0.0055. Problem: What conclusion would you make at each of the following significance levels? State the conclusion in context. (a) $\alpha = 0.01$		
	(b) $\alpha = 0.001$		
9.2	Significance tests for population proportions		
9.2A	Significance tests for population proportions		
	Conditions must be met:		
	Otherwise we can't infer to the population or establish cause and effect		
	• • • • • • • • • • • • • • • •		
	2. Normal: sampling distribution of the statistic is		
	Normal condition for proportions:		
	P will be replaced with \mathbf{p}_0 which is the		
	3. Independent: sampling with replacement for the population allows us to use standard deviation formulas, or if sampling without replacement, we meet the 10% condition for independence		
	· · · · · · · · · · · · · · · · · · ·		
9.2A	Calculations: Test statistic and P-value		
	A significance test uses sample data to measure the strength of evidence against H _o . Here are some principles that apply to most tests:		
	The test compares a calculated from sample data with the value of the		
	parameter stated by the		
	 Values of the statistic far from the in the direction specified by the 		
	alternative hypothesis give		
	I o assess now far the statistic is from the parameter, the statistic. This value is called the test statistic:		
	The test statistic measures how far the sample result is from the, in		
	what, on a standardized scale.		
	You can use the test statistic to find the of the test		

9.2A	Four step process for significance testing for proportions		
	State: What hypotheses do you want to test, and at what significance level?		
	 Test the hypothesis that true proportion of(context) is _(p₀) at a 		
	significance level.		
	• $H_0: p = p_0$		
	• $H_a: p > p_0 \text{ or } p < p_0 \text{ or } p \neq p_0$		
	 Where p is the true proportion of(context) 		
	Plan: Name procedure you are using. Check conditions.		
	Use a one-sample z-test for proportions		
	 Random condition? Random Sample or Random Assignment 		
	• Normal condition? $np_0 \ge 10$ and $n(1-p_0) \ge 10$		
	 Independence condition? independent or 10% rule 		
	Do: If conditions are met perform calculations		
	Compute <i>n</i>		
	• Compute the test statistic z (show work) $z = \frac{p - p_0}{z}$		
	$\int \underline{p_0(1-p_0)}$		
	$\bigvee n$		
	• Find the P-value = normal cdf(lower bound, upper bound, $\mu = 0$, $\sigma = 1$)		
	If Ha is > then lower bound is test statistic, upper bound is 10		
	If Ha is < then lower bound is -10, upper bound is test statistic		
	If Ha is \neq then use one of the above and times by 2.		
	Conclude: Interpret the results of your test in the context of the problem		
	Since our P-value is greater than our significance level we fail		
	to reject H_0 . We do not have sufficient evidence $(H_2$ in		
	context)		
	 Since our P-value is less than our significance level we reject 		
	H_0 . We have sufficient evidence $(H_2 \text{ in context})$		
9.2A	What happens when the data don't support Ha?		
	. If you aren't paying attention, you may the test. The		
	test will give you the, fail to reject H _o . A lot more work with the		
	!		

9.2A	Sample: On shows like American Idol, contestants often wonder if there is an advantage to performing last. To investigate this, a random sample of 600 American Idol fans is selected and they are shown
	the audition tapes of 12 never-before-seen contestants. For each fan, the order of the 12 videos is randomly determined. Thus, if the order of performance doesn't matter, we would expect
	approximately 1/12 of the fans to prefer the last contestant they view. In this study, 59 of the 600 fans
	advantage to going last?
	State: Test the hypothesis that true proportion of
	H_0 :
	H_a :
	where <i>p</i> = the true proportion of
	Plan: If conditions are met, we will perform a Random:
	Normal:
	Independent:
	Do:
	Conclude: Since the <i>P</i> -value is than, we,
	the null hypothesis. There convincing evidence to conclude that there is an advantage to performing last in American Idol.
9.2A	Sample: According to the National Campaign to Prevent Teen and Unplanned Pregnancy, 20% of
	themselves. The counselor at a large high school worries that the actual figure might be higher at her
	school. To find out, she gives an anonymous survey to a random sample of 250 of the school's 2800 students. All 250 respond and 63 admit to sending or posting sexual images. Carry out a significance
	test at the α = 0.05 significance level. What conclusion should the counselor draw?
	State: Test the hypothesis that the true proportion of teens aged 13 to 19 that have electronically sent
	or posted sexually suggestive images of themselves is 0.20
	H_0 : p = 0.2 H_a : p > 0.2 where p is the proportion of students "sexting" Plan: We will perform a one-sided z-test for proportions
	Random?
	Normal?
	Independent?
	P-value
	Conclude: Since the <i>P</i> -value is than, we
	the null hypothesis. There convincing evidence to conclude that more than 20% of teens
	at the concerned controlating explicit protes of themselves.

0.20	More on Two sided tests
	We perform a two sided test when looking for convincing evidence that the true parameter is different from the hypothesized value of the parameter,
	Sample: According to the Centers for Disease Control and Prevention (CDC) Web site, 50% of high school students have never smoked a cigarette. Taya wonders whether this national result hold true in her large, urban high school. For her statistics class, Taya takes an SRS of 150 students from her school. She gets responses from all 150 students, and 90 say that they have never smoked a cigarette. What should Taya conclude? Give appropriate evidencd to support your answer. State: Test the hypothesis that true proportion of
	H_{a} :
	H .
	where $p =$ the true proportion of
	 Plan: If conditions are met, we will perform a Random:
	Normal:
	Independent:
	Do:
	Conclude: Since the P-value is than, we,
	Conclude: Since the <i>P</i> -value is than, we, the null hypothesis. There convincing evidence to conclude that
9 2B	Conclude: Since the <i>P</i> -value is than, we the null hypothesis. There convincing evidence to conclude that
9.2B	Conclude: Since the P-value is than, we the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire
9.2B	Conclude: Since the <i>P</i> -value is than, we, the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion.
9.2B	Conclude: Since the P-value is than, we, we, the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the may be. Therefore, a confidence interval can be more
9.2B	Conclude: Since the P-value is than, we, we the null hypothesis. There convincing evidence to conclude that <u>Confidence intervals and significance tests</u> Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the Confidence interval can be more
9.2B	Conclude: Since the P-value is than, we the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the Sample: Taya found that 90 of an SRS of 150 students said that they had never smoked a cigarette. We checked the conditions for performing the significance test earlier. Before we construct a
9.2B	Conclude: Since the <i>P</i> -value is than, we the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the Confidence intervals give us an idea of what the Sample: Taya found that 90 of an SRS of 150 students said that they had never smoked a cigarette. We checked the conditions for performing the significance test earlier. Before we construct a confidence interval for the population proportion p, we should check that both $n p \ge 10$ and
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9.2B	Conclude: Since the <i>P</i> -value is than, we the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the may be. Therefore, a confidence interval can be more Sample: Taya found that 90 of an SRS of 150 students said that they had never smoked a cigarette. We checked the conditions for performing the significance test earlier. Before we construct a confidence interval for the population proportion p, we should check that both $np \ge 10$ and $n(1-p) \ge 10$. Since the number of successes and the number of failures in the sample are 90 and 60, we can proceed with our calculations. Calculate the 95% confidence interval for the true proportion of students at the school that never smoked a cigarette. State: Estimate the true proportion of students at the high school that have
9.2B	Conclude: Since the <i>P</i> -value is than, we the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the may be. Therefore, a confidence interval can be more Sample: Taya found that 90 of an SRS of 150 students said that they had never smoked a cigarette. We checked the conditions for performing the significance test earlier. Before we construct a confidence interval for the population proportion p, we should check that both $np \ge 10$ and $n(1-p) \ge 10$. Since the number of successes and the number of failures in the sample are 90 and 60, we can proceed with our calculations. Calculate the 95% confidence interval for the true proportion of students at the school that never smoked a cigarette. State: Estimate the true proportion of students at the high school that have
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9.2B	Conclude: Since the <i>P</i> -value is than, we the null hypothesis. There convincing evidence to conclude that Confidence intervals and significance tests Significance tests can tell us if the smaller specific population proportion from the entire population proportion. Confidence intervals give us an idea of what the may be. Therefore, a confidence interval can be more Sample: Taya found that 90 of an SRS of 150 students said that they had never smoked a cigarette. We checked the conditions for performing the significance test earlier. Before we construct a confidence interval for the population proportion p, we should check that both $np \ge 10$ and $n(1-p) \ge 10$. Since the number of successes and the number of failures in the sample are 90 and 60, we can proceed with our calculations. Calculate the 95% confidence interval for the true proportion of students at the school that never smoked a cigarette. State: Estimate the true proportion of students. Conditions were checked in the significance test above. Do: $z^* =$ Interval:

9.2B	Connection between confidence intervals and two sided significance tests There is a link between confidence intervals and two-sided significance tests:		
	 The confidence interval gives an approximate range of values p_o that would 		
	by a two sided test at the	s	ignificance level.
	However, with proportions the link is		because of the
	standard error used for the confidence interval is based on the sample proportion p , while the		
	denominator statistic is based on the value		
	Test statistic: $z = \frac{p - p_o}{\sqrt{\frac{p_o(1 - p_o)}{n}}}$ Confidence interval: $p \pm z * \sqrt{\frac{p(1 - p)}{n}}$		
	 So on the AP exam it is acceptable to u address a alternati 	se a confidence interval rat ve hypotheses but	her than the test statistic to alternative hypotheses.
9.1B	Errors in a	Significance Te	st
9.1B	Making Conclusions based on the sample		
	When we draw conclusions from a significance	test, we our co	
		Truth about	the population
		H ₀ is true	H_0 is false (H_a true)
	Your conclusion Reject H _o		
	sample Fail to reject H _o		
9.18	What happens when our conclusions are 1. Type 1 error: (Determined they are guilty when really 2. Type 2 error: (Determined they are innocent when really Type I But you should have found this Sample: Heading into the election Lack needs	There are two types they are innocent) ally they are guilty) sund s Type II But you s have four You found this	s of mistakes we can make:
9.10	feeling pretty confident that he will win the close	tion Is his confidence war	$H_0: p = 0.5$
	receiping pretty confident that ne will win the election. Is his confidence warranted? $H_a: p > 0.5$		
	a.) Describe in context a type 1 error and it	s consequences.	
	b.) Describe in context a type 2 error and it	s consequences.	



9.1B	A certain cigarette brand advertises that the mean r	icotine content of their cigarettes is 1.5 mg, but you		
	are suspicious and plan to investigate the advertise	d claim by testing the hypotheses $H_0: \mu = 1.5$		
	versus H_a : $\mu > 1.5$ at the $\alpha = 0.05$ significance level. You will do so by measuring the nicotine content			
	of 30 randomly selected cigarettes of this brand.			
	(a) Describe what a Type I error would be in this	s context.		
	(b) Describe what a Type II error would be in this context.			
	(a) Energy the mean estimated and the baselike order at the transmission of transmission of the transmission of transmissi			
	(c) From the perspective of public health, which Explain	endition is the sender of the		
	(d) You have determined that at the $\alpha = 0.05$ sid	inificance level, the power of the test against the		
	alternative $\mu = 1.75$ is 0.88. Explain what the	e power of the test means in the context of the		
	problem.			
9 1B	How is power affected by different parts of your	study?		
0.10	http://digitalfirst.bfwpub.com/stats_applet/stats_applet_9_power.html			
	Higher power	Lower power		
	α (significance level)	α (significance level)		
	sample size	sample size		
	• standard deviation	• standard deviation		
	(variance/spread)	(variance/spread)		
	 Ine difference between H₀ value and the true population value 	 Ine difference between H₀ value and the true population value 		
	Use a -sided test	Use a -sided test		
9.1B	Sample: The manager of a fast-food restaurant war	nts to reduce the proportion of drive-through		
	customers who have to wait more than 2 minutes to	receive their food once their order is placed.		
	Based on store records, the proportion of customers	s who had to wait at least 2 minutes was $p = 0.63$.		
	orders. During the pext month the manager assigns and	ect a random sample of drive-through times and		
	test the following hypotheses:			
	H_{a} : $p = 0.63$ where $p =$ the true proportion of drive-through customers who have			
	H : p < 0.63 to wait more than 2 minutes after their order is placed to receive their food.			
	Suppose that the manager of the fast food restaura	ht wants to change some aspects of his study. How		
	will these changes affect the power of the test?			
	1. To reduce the possibility of a Type I error	and avoid the possibility of unnecessarily paying		
	an extra employee, the manager reduces the	e significance level from 0.10 to 0.01.		
	2. To justify the additional cost of the extra e	employee, the manager decides that the true		
	proportion must be reduced to at most 0.53.			
	3. To get faster results, the manager reduce	es the sample size from 250 to 100		

9.1B	Planning studies: Determining Sample Size		
	Here are the questions we must answer to decide how many observations we need:		
	1. Significance level? If you insist on a significance level (such as 1% rather		
	than 5%), you have to take asample. A smaller significance level requires		
	evidence to reject the null hypothesis.		
	2. Practical Importance ? At any significance level and desired power, detecting a		
	requires a than detecting a large difference.		
	3. Power ? If you insist on power (such as 99% rather than 90%), you will need a		
	sample. Higher power gives a better chance of detecting a		
9.3	Significance tests for population means		
9.3A	When working with population we use since they are		
	based from the z-values from a standard normal distribution.		
	When working with population, we use from a t distribution with		
	n – 1 degrees of freedom.		
	To keep it straight we have "zap tax" for		
9.3A	Conditions for population means using a t distribution		
	Random: The data come from a from a sample of size n from the		
	population of interest or a This condition is very important.		
	a Normal		
	 <u>Inormal</u>. Sample size less than 15: use t procedures if the data appear close to normal. If it is 		
	skewed or has outliers use t procedures.		
	Sample size of at least 15: the t procedures can be used except if the data has		
	Large samples: The targedures can be used even for clearly skewed distributions when the		
	Large samples: The t procedures can be used even for clearly skewed distributions when the sample is		
	• Independent: individual observations are independent so if we sample without replacement the		
	10% must be met		
9.3A	Calculations: Test statistic and P-value		
	• Perform calculations assuming the H _o		
	 The test statistic measures now from the parameter value specified by H_o in standardized units 		
	 Due to our t distribution our test statistic is 		
	vvnen the normal condition is met, this statistic has a with degrees		
	UTITEEUUTT.		
	 Once we have calculated the test statistic we can use table b of a calculator to find the		
9.3A	Using to calculator to compute P-values from a t distribution		
	Go the the distribution menu (2 nd vars)		
	Choose tcdf(
1			

Four step process for significance testing for means		
Stat • •	e: What hypotheses do you want to test, and at what significance level? Test the hypothesis that true mean of(context) is _(μ_0) at a significance level. $H_0: \mu = \mu_0$ $H_a: \mu > \mu_0$ or $\mu < \mu_0$ or $\mu \neq \mu_0$ Where μ is the true mean of(context).	
Plar • •	n: Name procedure you are using. Check conditions. Use a one-sample t-test for means Random condition? Random Sample or Random Assignment Normal condition? n<15, IF close to normal, NO outliers or skewness 15≤n<30, IF NO outliers or strong skewness n≥30 -meets normal condition no matter what- Independence condition? independent or 10% rule	
Do: •	If conditions are met, perform calculations Compute \bar{x} and s_x	
•	Compute the test statistic t (show work) with df = n -1, $t = \frac{\overline{x} - \mu_0}{S_x / \sqrt{n}}$	
•	Find the P-value = tcdf (lower bound, upper bound, df) If Ha is > then lower bound is test statistic, upper bound is 100 If Ha is < then lower bound is -100, upper bound is test statistic If Ha is \neq then use one of the above and times by 2.	
Con •	clude: Interpret the results of your test in the context of the problem. Since our P-value is greater than our significance level, we fail to reject H ₀ . We do not have sufficient evidence(H _a in context)	

9.3A	Sample: Every see if drivers ob miles per hour, o results:	road ha ey thes or mph	as one a se lower) of a ra	at some speed ndom s	e point— limits, a sample o	-constru a police of 10 dr	uction zo officer ivers in	ones tha used a a 25 m	at have radar g ph cons	much I un to m struction	ower sp leasure n zone.	beed limit the spee Here are	s. To d (in e the
		27	33	32	21	30	30	29	25	27	34		
	Problem: Can withe posted 25 m	ve con iph spe	clude th ed limit	at the a ?	average	speed	of drive	rs in thi	s consti	ruction	zone is	greater t	han
	<i>State:</i> We will te zone.	est the H₀:	hypothe µ =	esis tha I	t the tru H _a : μ	e mean	speed	of drive	rs is 25	mph in	this co	nstructior	ı
	Where μ is the _												
	<i>Plan:</i> We will pe Random: Normal:	erform	a										
	Independent: Do:												
	Conclude: Sinc	e the <i>l</i>	2-value i	S		than			w	/e			
	the null hypothe	sis. Th	nere		con	wincing	eviden	ce to co	nclude	that			
	(b) Given your c have made? Ex	onclus plain v	ion in pa vhat this	art (a), [,] s mistak	which ki ke mear	ind of m	nistake- s conte	 _a Type	elora	Type II	error—	could you	L

A	Sample: A college professor suspects that students at his school are getting less than 8 hours of sleep a night, on average. To test his belief, the professor asks a random sample of 28 students, "how much sleep did you get last night?" Here are the data (in hours):																										
	9	68	6	88	6	6.5	67	9	4	3	4	56	11	63	6	6	1	0 7	7	8	4.5	ę	97	7	7		
	Do sig	thes nifica	e da ance	ata pro test a	ovid at th	e con e α =	vincir 0.05	ng e leve	vide el to	nce hel	e in p a	supp answe	ort c r thi	of the s que	pro stio	fess n.	SO	rs sı	JSP	oici	on?	C	Carr	y c	out	а	
	Sta Wh	ate: T iere i	est⊺ H₀: uis⊺	he hy μ = _ the tru	potl 	nesis nean	that t I hours	he t Ha: µ of :	rue i i sleep	nea o a	an col	hours	s of s	leep (ent ge	colle ets.	ege	e Si	tude	nt	ge	ts is	8	8 ho	urs	6.		
	Pla Ra No	n <i>n:</i> W ndon rmal	/e w n? S ? Th	ill per RS o e nor	forn f 28 mal	n a or stude proba	ne-sai ents a ability	mple it th plo	e t-te e col t shc	st f leg ws	or e	mear	IS	0													
	Ind Do	eper :	nder	t?																							
	Co the	<i>nclu</i> null	de: hyp	Since	the	<i>P-</i> va	llue is	·		_ c	on	than . vincin	g ev	idenc	e to) CO	onc	, clude	we e th	nat							
																·											

In the child Each child cubes don'			3	Two sided significance tests								
Each child	In the children's game Don't Break the Ice, small plastic ice cubes are squeezed into a square frame. Each child takes turns tapping out a cube of "ice" with a plastic hammer hoping that the remaining cubes don't collapse. For the game to work correctly, the cubes must be big enough so that they hold											
cubes don'												
CIDES OOD												
apple other in place in the placetic frame but not ap hig that they are too difficult to top out. The machine												
thet reach other in place in the plastic frame but not so big that they are too difficult to tap out. The machine												
that produces the plastic ice cubes is designed to make cubes that are 29.5 millimeters (mm) wide, but the actual width varies a little. To make sure the machine is working well, a supervisor inspects a												
										random sa	mple of s	50 cubes e
from a sam	ple take	en during o	ne hour.									
		-										
variable	Ν	Mean	SEmean	stdDev	Min	Q1	Median	Q3	Max			
Width	50	29 4874	0.0132	0.0935	29 2717	29 4225	29 4821	29 5544	29 7148			
Do these o	tata dive		a evidence	that the m	ean width	of cubes n	roduced th	is hour is r	2017 1 10			
	iala give		y evidence			or cubes p			101 29.5			
11111 ?												
			_									
State: Test	the hype	othesis tha	t the true m	nean width	of "ice cub	pes" is 29.5	mm.					
H _c): μ = 29	.5mm	H _a : μ	≠ 29.5mm								
Where µ is	the true	e mean wid	th of "ice cເ	ubes".								
Plan: We v	vill perfo	rm a one-s	sample t-tes	st for mear	าร							
Random? S	SRS of		ice cubes"									
Normal?												
Independer	nt? More	than	"ice	o cubes" pr	oduced							
			ICC	cubes pi	ouuceu.							
<i>D</i> 0.												
Conclude:	Conclude: Since the P-value is than											
the null by	Since th	he <i>P-</i> value	is	than			, we					
	Since ti oothesis.	he <i>P-</i> value . There	is	than _	a evidence	e to conclu	, we de that					
	Since ti oothesis.	he <i>P-</i> value . There	is	than _ _ convincin	g evidence	e to conclu	, we de that					
	Since ti oothesis.	he <i>P-</i> value . There	is	than _ _ convincin	g evidence	e to conclu	, we de that					
	Since ti oothesis.	he <i>P</i> -value . There	is	than convincin	g evidence	e to conclu	, we de that					
Two sided	Since the sist	he <i>P</i> -value . There	is	than _ convincin	g evidence	e to conclue	, we de that		that we			
Two sided • Unf	since the sist	he <i>P</i> -value . There cance test	is s and conf	than _ _ convincin fidence int	g evidence t <u>ervals</u> ell us the _	e to conclue	, we de that	, for	that we			
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Two sided • Unf nee • The	since the sis.	he <i>P</i> -value . There cance test ely the sign fidence inte	is s and conf ificance tes erval. en significa	than _ convincin fidence int st doesn't te ance tests a	g evidence t <u>ervals</u> ell us the _ and confide	e to conclue 	, we de that als is even	, for	that we			
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	the actual v random sau from a sam variable Width Do these of mm? State:Test Ho Where μ is Plan: We v Random? S Normal? Independer Do:	the actual width variandom sample of from a sample take from a sample take variable N Width 50 Do these data give mm? State: Test the hype $H_0: \mu = 29$ Where μ is the true Plan: We will perfor Random? SRS of _ Normal? Independent? More Do:	the actual width varies a little. random sample of 50 cubes e from a sample taken during o variable N Mean Width 50 29.4874 Do these data give convincin mm? State:Test the hypothesis tha $H_0: \mu = 29.5$ mm Where μ is the true mean wid Plan: We will perform a one-s Random? SRS of" Normal? Independent? More than Do:	the actual width varies a little. To make random sample of 50 cubes every hour a from a sample taken during one hour. variable N Mean SEmean Width 50 29.4874 0.0132 Do these data give convincing evidence mm? State: Test the hypothesis that the true n H_0 : $\mu = 29.5$ mm H_a: μ Where μ is the true mean width of "ice cu Random? SRS of "ice cubes" Normal? Independent? More than "ice Do:	the actual width varies a little. To make sure the m random sample of 50 cubes every hour and measu from a sample taken during one hour. $\begin{array}{r c c c c c c c c c c c c c c c c c c c$	the actual width varies a little. To make sure the machine is variandom sample of 50 cubes every hour and measures their ware from a sample taken during one hour. variable N Mean SEmean stdDev Min Width 50 29.4874 0.0132 0.0935 29.2717 Do these data give convincing evidence that the mean width mm? State: Test the hypothesis that the true mean width of "ice cubes" Ho: $\mu = 29.5$ mm Ha: $\mu \neq 29.5$ mm Where μ is the true mean width of "ice cubes". Plan: We will perform a one-sample t-test for means Random? SRS of "ice cubes" "ice cubes" produced. Do: Do: Do: "ice cubes" produced.	the actual width varies a little. To make sure the machine is working well random sample of 50 cubes every hour and measures their width. The from a sample taken during one hour. variable N Mean SEmean stdDev Min Q1 Width 50 29.4874 0.0132 0.0935 29.2717 29.4225 Do these data give convincing evidence that the mean width of cubes p mm? State:Test the hypothesis that the true mean width of "ice cubes" is 29.5 H ₀ : μ = 29.5mm H _a : $\mu \neq 29.5mm$ Where μ is the true mean width of "ice cubes". Plan: We will perform a one-sample t-test for means Random? SRS of "ice cubes" Normal? Independent? More than "ice cubes" produced. Do: Do:	the actual width varies a little. To make sure the machine is working well, a superv random sample of 50 cubes every hour and measures their width. The output sun from a sample taken during one hour. variable N Mean SEmean stdDev Min Q1 Median Width 50 29.4874 0.0132 0.0935 29.2717 29.4225 29.4821 Do these data give convincing evidence that the mean width of cubes produced th mm? State: Test the hypothesis that the true mean width of "ice cubes" is 29.5mm. H ₀ : $\mu = 29.5mm$ H _a : $\mu \neq 29.5mm$ Where μ is the true mean width of "ice cubes". Plan: We will perform a one-sample t-test for means Random? SRS of "ice cubes" "ice cubes" produced. Do: Do: Do: "ice cubes" produced.	the actual width varies a little. To make sure the machine is working well, a supervisor inspect random sample of 50 cubes every hour and measures their width. The output summarizes the from a sample taken during one hour. variable N Mean SEmean stdDev Min Q1 Median Q3 Width 50 29.4874 0.0132 0.0935 29.2717 29.4225 29.4821 29.5544 Do these data give convincing evidence that the mean width of cubes produced this hour is r mm? State: Test the hypothesis that the true mean width of "ice cubes" is 29.5mm. H_0: $\mu = 29.5mm$ H_a: $\mu \neq 29.5mm$ Where μ is the true mean width of "ice cubes". Plan: We will perform a one-sample t-test for means Random? SRS of "ice cubes" Normal? Independent? More than "ice cubes" produced. Do:			

9.3B	Inference for m Compara	eans: Paired da tive studies are	<u>ata</u> more convincing than _	investigations.					
	Therefore	e, one-sample ir	iference	than comparative inference.					
	 Study designs that involve making on the same individual, or one observation on each of individuals, result in paired data. (this experiment design is called matched pairs) When paired data result from measuring the same variable twice, we can make comparisons by analyzing in each pair. If the conditions for inferences are met, we can use one-sample t procedures to perform inference about the These methods are called paired t procedures. 								
9.3B	Sample: For the which line was fa they randomly se However, one of each of them wo used the regular They entered the	ir second seme aster in the supe elected 15 times them used the uld use, they flip lane. If it was the eir randomly ass	ster project in AP Statis ermarket, the express la during a week, went to express lane and the or oped a coin. If it was he ails, Libby used the reg igned lanes at the sam	tics, Libby and Kathryn decided to investigate ane or the regular lane. To collect their data, the same store, and bought the same item. ther used a regular lane. To decide which lane eads, Libby used the express lane and Kathryn ular lane and Kathryn used the express lane. e time and each recorded the time in seconds it					
	took them to con Time in express lane	nplete the transa Time in regular lane	action.						
	(seconds)	(seconds)							
	337	342							
	226	472							
	502	456							
	408	529							
	151	181							
	284	339							
	150	229							
	357	263							
	349	332							
	257	352							
	321	341							
	383	397							
	565	694							
	363	324							
	85	127							
	Carry out a test t State: Plan:	to see if there is	convincing evidence th	hat the express lane is faster.					

	Do:
	Conclude:
9.3B	About naired data
3.50	Individual scores are
	 Differences in scores are not dependent (
	Be sure to report
	 If subjects in an experiment were not randomly chosen, we can't generalize our findings to
	 If subjects in an experiment were not randomly assigned a treatment, we can't make an
	· · · · · · · · · · · · · · · · · · ·
	A confidence interval gives than a significance test.
9.3B	Using tests wisely
	Significance tests are widely used in reporting the in many fields.
	New drugs require significant evidence of
	Courts ask about in hearing discrimination cases.
	In all cases, statistical significance is valued because it points to an effect that is
	 Statistical significance is not the same thing as Pay attention to the statistical significance to sucid this
	actual data () along with the statistical significance to avoid this.
	 The foolish user of statistics who feeds the data to a calculator or computer without exploratory
	analysis will often be embarassed. If you were to actually, would it be
	worthwhile to calculate?
	 Don't ignore the lack of significance. Usually will be needed to help us
	make some conclusions.
	 When planning a study, verify that the test you plan to use has a
	of detecting a difference of the size you hope to find
	Statistical informacional valid for
	• Statistical interence is not valid for the set how the date was
	experiments will yield invalid results. Always ask now the data was
	Running tests multiple times to get the significance you want will have