

# Chapter 3

## Fundamental Sampling Distributions

Department of Statistics and Operations Research



October 15, 2019

- 1 Random sampling and statistics
- 2 Sampling Distribution of Means and the Central Limit Theorem
- 3 Sampling Distribution of the Difference between Two Means
- 4 Sampling Distribution of the Variance
- 5 The Student's Distribution
- 6 The Fisher Distribution
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## Definitions

- 1 A population consists of the totality of the observations with which we are concerned.
- 2 A sample is a subset of a population.
- 3 Any function of the random variables constituting a random sample is called a statistic.

- Sample mean:  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$ .

- Sample median:  $\tilde{X} = \begin{cases} x_{\frac{n+1}{2}}, & \text{if } n \text{ is odd,} \\ \frac{1}{2}(x_{\frac{n}{2}} + x_{\frac{n}{2}+1}), & \text{if } n \text{ is even.} \end{cases}$

- Sample variance:  $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$ .

The computed value of  $S^2$  for a given sample is denoted by  $s^2$ .

### Theorem

If  $S^2$  is the variance of a random sample of size  $n$ , we may write

$$S^2 = \frac{1}{n-1} \left[ \sum_{i=1}^n X_i^2 - n\bar{X}^2 \right]$$

- Sample standard deviation:  $S = \sqrt{S^2}$

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## Definition

The probability distribution of a statistic is called a sampling distribution.

### Theorem

If  $X_1, X_2, \dots, X_n$  are independent random variables having normal distributions with means  $\mu_1, \mu_2, \dots, \mu_n$  and variances  $\sigma_1^2, \sigma_2^2, \dots, \sigma_n^2$ , respectively, then the random variable  $Y = a_1X_1 + a_2X_2 + \dots + a_nX_n$  has a normal distribution with mean

$$\mu_Y = a_1\mu_1 + a_2\mu_2 + \dots + a_n\mu_n$$

and variance

$$\sigma_Y^2 = a_1^2\sigma_1^2 + a_2^2\sigma_2^2 + \dots + a_n^2\sigma_n^2$$

Suppose that a random sample of  $n$  observations is taken from a normal population with mean  $\mu$  and variance  $\sigma^2$ . Each observation  $X_i$ ,  $i = 1, 2, \dots, n$ , of the random sample will then have the same normal distribution. Hence, we conclude that

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

has a normal distribution with mean

$$\mu_{\bar{X}} = \frac{1}{n} \{ \mu + \mu + \dots + \mu \} = \frac{1}{n} \sum_{i=1}^n \mu = \mu$$

and variance

$$\sigma_{\bar{X}}^2 = \frac{1}{n^2} \{ \sigma^2 + \sigma^2 + \dots + \sigma^2 \} = \frac{1}{n^2} \sum_{i=1}^n \sigma^2 = \frac{\sigma^2}{n}.$$



### Corollary

If  $X_1, X_2, \dots, X_n$  are independent random variables having normal distributions with means  $\mu$  and variances  $\sigma^2$ , then the sample mean  $\bar{X}$  is normally distributed with mean equal to  $\mu$  and standard deviation equal to  $\sigma/\sqrt{n}$ . Consequently the random variable

$$Z = \frac{(\bar{X} - \mu)}{\sigma/\sqrt{n}}$$

is a standard normal distribution.

### Theorem (Central limit theorem)

If  $\bar{X}$  is the mean of a random sample of size  $n$  taken from a population with mean  $\mu$  and finite variance  $\sigma^2$ , then the limiting form of the distribution of

$$Z = \frac{(\bar{X} - \mu)}{\sigma/\sqrt{n}}$$

as  $n \rightarrow \infty$ , is the standard normal distribution  $N(0, 1)$ .

The normal approximation for  $\bar{X}$  will generally be good if  $n \geq 30$ .

### Example

An electrical firm manufactures light bulbs that have a length of life that is approximately normally distributed, with mean equal to 800 hours and a standard deviation of 40 hours. Find the probability that a random sample of 16 bulbs will have an average life of less than 775 hours.

## Solution

Here  $\mu = 800$ ,  $\sigma = 40$  and  $n = 16$ . The random variable  $\bar{X}$  is normally distributed with mean  $\mu_{\bar{X}} = \mu = 800$  and standard deviation  $\sigma_{\bar{X}} = \sigma_X / \sqrt{n} = 10$ .

Then  $(\bar{X} - 800)/10$  is a standard normal distribution  $N(0, 1)$ .  
Hence,

$$P(\bar{X} < 775) = P\left(\frac{\bar{X} - 800}{10} < \frac{775 - 800}{10}\right) = P(Z < -2.5) = 0.0062.$$

### Example

Traveling between two campuses of a university in a city via shuttle bus takes, on average, 28 minutes with a standard deviation of 5 minutes. In a given week, a bus transported passengers 40 times. What is the probability that the average transport time was more than 30 minutes?

## Solution

In this case,  $\mu = 28$  and  $\sigma = 3$ . We need to calculate the probability  $Pr(\bar{X} > 30)$  with  $n = 40$ . Hence,

$$\begin{aligned} P(\bar{X} > 30) &= P\left(\frac{\bar{X} - 28}{5/\sqrt{40}} \geq \frac{30 - 28}{5/\sqrt{40}}\right) \\ &= P(Z \geq 2.53) = 1 - P(Z \leq 2.53) \\ &= 1 - 0.9925 = 0.0075. \end{aligned}$$

There is only a slight chance that the average time of one bus trip will exceed 30 minutes.

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## Theorem

If independent samples of size  $n_1$  and  $n_2$  are drawn at random from two populations, discrete or continuous, with means  $\mu_1$  and  $\mu_2$  and variances  $\sigma_1^2$  and  $\sigma_2^2$ , respectively, then the sampling distribution of the differences of means,  $\bar{X}_1 - \bar{X}_2$ , is approximately normally distributed with mean and variance given by

$$\mu_{\bar{X}_1 - \bar{X}_2} = \mu_1 - \mu_2 \text{ and } \sigma_{\bar{X}_1 - \bar{X}_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

Hence,

$$Z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\sigma_1^2/n_1 + \sigma_2^2/n_2}}$$

is approximately a standard normal variable.

If both  $n_1$  and  $n_2$  are greater than or equal to 30, the normal approximation for the distribution of  $\bar{X}_1 - \bar{X}_2$  is good.



### Example

Eighteen specimens are painted using type A, and the drying time, in hours, is recorded for each. The same is done with type B. The population standard deviations are both known to be 1.0.

Assuming that the mean drying time is equal for the two types of paint, find  $P(\bar{X}_A - \bar{X}_B > 1.0)$ , where  $\bar{X}_A$  and  $\bar{X}_B$  are average drying times for samples of size  $n_A = n_B = 18$ .

## Solution

From the sampling distribution of  $\bar{X}_A - \bar{X}_B$ , we know that the distribution is approximately normal with mean

$\mu_{\bar{X}_A - \bar{X}_B} = \mu_A - \mu_B = 0$  and variance  $\sigma_{\bar{X}_A - \bar{X}_B}^2 = \frac{\sigma_A^2}{n_A} + \frac{\sigma_B^2}{n_B} = 1/9$ .

Corresponding to the value  $\bar{X}_A - \bar{X}_B = 1.0$ , we have

$$z = \frac{1 - (\mu_A - \mu_B)}{\sqrt{1/9}} = \frac{1 - 0}{\sqrt{1/9}} = 3$$

so

$$\Pr(Z > 3.0) = 1 - P(Z < 3.0) = 1 - 0.9987 = 0.0013.$$

### Example

The television picture tubes of manufacturer  $A$  have a mean lifetime of 6.5 years and a standard deviation of 0.9 year, while those of manufacturer  $B$  have a mean lifetime of 6.0 years and a standard deviation of 0.8 year. What is the probability that a random sample of 36 tubes from manufacturer  $A$  will have a mean lifetime that is at least 1 year more than the mean lifetime of a sample of 49 tubes from manufacturer  $B$ ?

## Solution

We are given the following information:

Population 1	Population 2
$\mu_1 = 6.5$	$\mu_2 = 6.0$
$\sigma_1 = 0.9$	$\sigma_2 = 0.8$
$n_1 = 36$	$n_2 = 49$

If we use, the sampling distribution of  $\bar{X}_1 - \bar{X}_2$  will be approximately normal and will have a mean and standard deviation

$$\mu_{\bar{X}_1 - \bar{X}_2} = 6.5 - 6.0 \text{ and } \sigma_{\bar{X}_1 - \bar{X}_2} = \sqrt{\frac{0.81}{36} + \frac{0.64}{49}} = 0.189$$

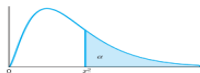
Hence,

$$\begin{aligned} \Pr(\bar{X}_1 - \bar{X}_2 \geq 1.0) &= P(Z > 2.65) = 1 - P(Z < 2.65) \\ &= 1 - 0.9960 = 0.0040. \end{aligned}$$

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### Theorem

- 1 If  $X_1, X_2, \dots, X_n$  an independent random sample that have the same standard normal distribution then  $X = \sum_{i=1}^n X_i^2$  is chi-squared distribution, with degrees of freedom  $\nu = n$ .
- 2 The mean and variance of the chi-squared distribution  $\chi^2$  with  $\nu$  degrees of freedom are  $\mu = \nu$  and  $\sigma^2 = 2\nu$ .



**Table A.5** Critical Values of the Chi-Squared Distribution

$v$	$\alpha$									
	0.995	0.99	0.98	0.975	0.95	0.90	0.80	0.75	0.70	0.50
1	0.0 <sup>4</sup> 393	0.0 <sup>3</sup> 157	0.0 <sup>3</sup> 628	0.0 <sup>3</sup> 982	0.00393	0.0158	0.0642	0.102	0.148	0.455
2	0.0100	0.0201	0.0404	0.0506	0.103	0.211	0.446	0.575	0.713	1.386
3	0.0717	0.115	0.185	0.216	0.352	0.584	1.005	1.213	1.424	2.366
4	0.207	0.297	0.429	0.484	0.711	1.064	1.649	1.923	2.195	3.357
5	0.412	0.554	0.752	0.831	1.145	1.610	2.343	2.675	3.000	4.351
6	0.676	0.872	1.134	1.237	1.635	2.204	3.070	3.455	3.828	5.348
7	0.989	1.239	1.564	1.690	2.167	2.833	3.822	4.255	4.671	6.346
8	1.344	1.647	2.032	2.180	2.733	3.490	4.594	5.071	5.527	7.344
9	1.735	2.088	2.532	2.700	3.325	4.168	5.380	5.899	6.393	8.343
10	2.156	2.558	3.059	3.247	3.940	4.865	6.179	6.737	7.267	9.342
11	2.603	3.053	3.609	3.816	4.575	5.578	6.989	7.584	8.148	10.341
12	3.074	3.571	4.178	4.404	5.226	6.304	7.807	8.438	9.034	11.340
13	3.565	4.107	4.765	5.009	5.892	7.041	8.634	9.299	9.926	12.340
14	4.075	4.660	5.368	5.629	6.571	7.790	9.467	10.165	10.821	13.339
15	4.601	5.229	5.985	6.262	7.261	8.547	10.307	11.037	11.721	14.339
16	5.142	5.812	6.614	6.908	7.962	9.312	11.152	11.912	12.624	15.338
17	5.697	6.408	7.255	7.564	8.672	10.085	12.002	12.792	13.531	16.338
18	6.265	7.015	7.906	8.231	9.390	10.865	12.857	13.675	14.440	17.338
19	6.844	7.633	8.567	8.907	10.117	11.651	13.716	14.562	15.352	18.338
20	7.434	8.260	9.237	9.591	10.851	12.443	14.578	15.452	16.266	19.337
21	8.034	8.897	9.915	10.283	11.591	13.240	15.445	16.344	17.182	20.337
22	8.643	9.542	10.600	10.982	12.338	14.041	16.314	17.240	18.101	21.337
23	9.260	10.196	11.293	11.689	13.091	14.848	17.187	18.137	19.021	22.337
24	9.886	10.856	11.992	12.401	13.848	15.659	18.062	19.037	19.943	23.337
25	10.520	11.524	12.697	13.120	14.611	16.473	18.940	19.939	20.867	24.337
26	11.160	12.198	13.409	13.844	15.379	17.292	19.820	20.843	21.792	25.336
27	11.808	12.878	14.125	14.573	16.151	18.114	20.703	21.749	22.719	26.336
28	12.461	13.565	14.847	15.308	16.928	18.939	21.588	22.657	23.647	27.336
29	13.121	14.256	15.574	16.047	17.708	19.768	22.475	23.567	24.577	28.336
30	13.787	14.953	16.306	16.791	18.493	20.599	23.364	24.478	25.508	29.336
40	20.707	22.164	23.838	24.433	26.509	29.051	32.345	33.66	34.872	39.335
50	27.991	29.707	31.664	32.357	34.764	37.689	41.449	42.942	44.313	49.335
60	35.534	37.485	39.699	40.482	43.188	46.459	50.641	52.294	53.809	59.335

Figure : Table A.5 Critical Values of the Chi-Squared Distribution

**Table A.5 (continued) Critical Values of the Chi-Squared Distribution**

<i>v</i>	$\alpha$									
	<b>0.30</b>	<b>0.25</b>	<b>0.20</b>	<b>0.10</b>	<b>0.05</b>	<b>0.025</b>	<b>0.02</b>	<b>0.01</b>	<b>0.005</b>	<b>0.001</b>
<b>1</b>	1.074	1.323	1.642	2.706	3.841	5.024	5.412	6.635	7.879	10.827
<b>2</b>	2.408	2.773	3.219	4.605	5.991	7.378	7.824	9.210	10.597	13.815
<b>3</b>	3.665	4.108	4.642	6.251	7.815	9.348	9.837	11.345	12.838	16.266
<b>4</b>	4.878	5.385	5.989	7.779	9.488	11.143	11.668	13.277	14.860	18.466
<b>5</b>	6.064	6.626	7.289	9.236	11.070	12.832	13.388	15.086	16.750	20.515
<b>6</b>	7.231	7.841	8.558	10.645	12.592	14.449	15.033	16.812	18.548	22.457
<b>7</b>	8.383	9.037	9.803	12.017	14.067	16.013	16.622	18.475	20.278	24.321
<b>8</b>	9.524	10.219	11.030	13.362	15.507	17.535	18.168	20.090	21.955	26.124
<b>9</b>	10.656	11.389	12.242	14.684	16.919	19.023	19.679	21.666	23.589	27.877
<b>10</b>	11.781	12.549	13.442	15.987	18.307	20.483	21.161	23.209	25.188	29.588
<b>11</b>	12.899	13.701	14.631	17.275	19.675	21.920	22.618	24.725	26.757	31.264
<b>12</b>	14.011	14.845	15.812	18.549	21.026	23.337	24.054	26.217	28.300	32.909
<b>13</b>	15.119	15.984	16.985	19.812	22.362	24.736	25.471	27.688	29.819	34.527
<b>14</b>	16.222	17.117	18.151	21.064	23.685	26.119	26.873	29.141	31.319	36.124
<b>15</b>	17.322	18.245	19.311	22.307	24.996	27.488	28.259	30.578	32.801	37.698
<b>16</b>	18.418	19.369	20.465	23.542	26.296	28.845	29.633	32.000	34.267	39.252
<b>17</b>	19.511	20.489	21.615	24.769	27.587	30.191	30.995	33.409	35.718	40.791
<b>18</b>	20.601	21.605	22.760	25.989	28.869	31.526	32.346	34.805	37.156	42.312
<b>19</b>	21.689	22.718	23.900	27.204	30.144	32.852	33.687	36.191	38.582	43.819
<b>20</b>	22.775	23.828	25.038	28.412	31.410	34.170	35.020	37.566	39.997	45.314
<b>21</b>	23.858	24.935	26.171	29.615	32.671	35.479	36.343	38.932	41.401	46.796
<b>22</b>	24.939	26.039	27.301	30.813	33.924	36.781	37.659	40.289	42.796	48.268
<b>23</b>	26.018	27.141	28.429	32.007	35.172	38.076	38.968	41.638	44.181	49.728
<b>24</b>	27.096	28.241	29.553	33.196	36.415	39.364	40.270	42.980	45.558	51.179
<b>25</b>	28.172	29.339	30.675	34.382	37.652	40.646	41.566	44.314	46.928	52.619
<b>26</b>	29.246	30.435	31.795	35.563	38.885	41.923	42.856	45.642	48.290	54.051
<b>27</b>	30.319	31.528	32.912	36.741	40.113	43.195	44.140	46.963	49.645	55.475
<b>28</b>	31.391	32.620	34.027	37.916	41.337	44.461	45.419	48.278	50.994	56.892
<b>29</b>	32.461	33.711	35.139	39.087	42.557	45.722	46.693	49.588	52.335	58.301
<b>30</b>	33.530	34.800	36.250	40.256	43.773	46.979	47.962	50.892	53.672	59.702
<b>40</b>	44.165	45.616	47.269	51.805	55.758	59.342	60.436	63.691	66.766	73.403
<b>50</b>	54.723	56.334	58.164	63.167	67.505	71.420	72.613	76.154	79.490	86.660
<b>60</b>	65.226	66.981	68.972	74.397	79.082	83.298	84.58	88.379	91.952	99.608



### Example

For a chi-squared distribution, find

- (a)  $\chi_{0.025}^2$  when  $\nu = 15$ ;
- (b)  $\chi_{0.01}^2$  when  $\nu = 7$ ;
- (c)  $\chi_{0.05}^2$  when  $\nu = 24$ .

### Solution

- (a) 27.488.
- (b) 18.475.
- (c) 36.415.

### Example

For a chi-squared distribution  $X$ , find  $\chi_{\alpha}^2$  such that

- (a)  $P(X > \chi_{\alpha}^2) = 0.99$  when  $\nu = 4$ ;
- (b)  $P(X > \chi_{\alpha}^2) = 0.025$  when  $\nu = 19$ ;
- (c)  $P(37.652 < X < \chi_{\alpha}^2) = 0.045$  when  $\nu = 25$ .

### Solution

- (a)  $\chi_{\alpha}^2 = \chi_{0.99}^2 = 0.297$ .
- (b)  $\chi_{\alpha}^2 = \chi_{0.025}^2 = 32.852$ .
- (c)  $\chi_{0.05}^2 = 37.652$ . Therefore,  $\alpha = 0.05 - 0.045 = 0.005$ .  
Hence,  $\chi_{\alpha}^2 = \chi_{0.005}^2 = 46.928$ .

### Theorem 23

If  $S^2$  is the variance of a random sample of size  $n$  taken from a normal population having the variance  $\sigma^2$ , then the statistic

$$\chi^2 = \frac{(n-1)S^2}{\sigma^2} = \sum_{i=1}^n \frac{(X_i - \bar{X})^2}{\sigma^2}$$

has a chi-squared distribution with  $\nu = n - 1$  degrees of freedom.

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### Theorem

Let  $Z$  be a standard normal random variable and  $V$  a chi-squared random variable with  $\nu$  degrees of freedom. If  $Z$  and  $V$  are independent, then the distribution of the random variable  $T$ , where

$$T = \frac{Z}{\sqrt{V/\nu}}$$

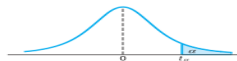
This is known as the  $t$ -distribution with  $\nu$  degrees of freedom.

### Corollary

Let  $X_1, X_2, \dots, X_n$  be independent random variables that are all normal with mean  $\mu$  and standard deviation  $\sigma$ . Let

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \quad \text{and} \quad S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

Then the random variable  $T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$  has a  $t$ -distribution with  $\nu = n - 1$  degrees of freedom.



**Table A.4 Critical Values of the  $t$ -Distribution**

$v$	$\alpha$						
	0.40	0.30	0.20	0.15	0.10	0.05	0.025
1	0.325	0.727	1.376	1.963	3.078	6.314	12.706
2	0.289	0.617	1.061	1.386	1.886	2.920	4.303
3	0.277	0.584	0.978	1.250	1.638	2.353	3.182
4	0.271	0.569	0.941	1.190	1.533	2.132	2.776
5	0.267	0.559	0.920	1.156	1.476	2.015	2.571
6	0.265	0.553	0.906	1.134	1.440	1.943	2.447
7	0.263	0.549	0.896	1.119	1.415	1.895	2.365
8	0.262	0.546	0.889	1.108	1.397	1.860	2.306
9	0.261	0.543	0.883	1.100	1.383	1.833	2.262
10	0.260	0.542	0.879	1.093	1.372	1.812	2.228
11	0.260	0.540	0.876	1.088	1.363	1.796	2.201
12	0.259	0.539	0.873	1.083	1.356	1.782	2.179
13	0.259	0.538	0.870	1.079	1.350	1.771	2.160
14	0.258	0.537	0.868	1.076	1.345	1.761	2.145
15	0.258	0.536	0.866	1.074	1.341	1.753	2.131
16	0.258	0.535	0.865	1.071	1.337	1.746	2.120
17	0.257	0.534	0.863	1.069	1.333	1.740	2.110
18	0.257	0.534	0.862	1.067	1.330	1.734	2.101
19	0.257	0.533	0.861	1.066	1.328	1.729	2.093
20	0.257	0.533	0.860	1.064	1.325	1.725	2.086
21	0.257	0.532	0.859	1.063	1.323	1.721	2.080
22	0.256	0.532	0.858	1.061	1.321	1.717	2.074
23	0.256	0.532	0.858	1.060	1.319	1.714	2.069
24	0.256	0.531	0.857	1.059	1.318	1.711	2.064
25	0.256	0.531	0.856	1.058	1.316	1.708	2.060
26	0.256	0.531	0.856	1.058	1.315	1.706	2.056
27	0.256	0.531	0.855	1.057	1.314	1.703	2.052
28	0.256	0.530	0.855	1.056	1.313	1.701	2.048
29	0.256	0.530	0.854	1.055	1.311	1.699	2.045
30	0.256	0.530	0.854	1.055	1.310	1.697	2.042
40	0.255	0.529	0.851	1.050	1.303	1.684	2.021
60	0.254	0.527	0.848	1.045	1.296	1.671	2.000
120	0.254	0.526	0.845	1.041	1.289	1.658	1.980
$\infty$	0.253	0.524	0.842	1.036	1.282	1.645	1.960

Figure : Table A.4 Critical Values of the  $t$ -Distribution

**Table A.4 (continued) Critical Values of the  $t$ -Distribution**

$\nu$	$\alpha$						
	<b>0.02</b>	<b>0.015</b>	<b>0.01</b>	<b>0.0075</b>	<b>0.005</b>	<b>0.0025</b>	<b>0.0005</b>
<b>1</b>	15.894	21.205	31.821	42.433	63.656	127.321	636.578
<b>2</b>	4.849	5.643	6.965	8.073	9.925	14.089	31.600
<b>3</b>	3.482	3.896	4.541	5.047	5.841	7.453	12.924
<b>4</b>	2.999	3.298	3.747	4.088	4.604	5.598	8.610
<b>5</b>	2.757	3.003	3.365	3.634	4.032	4.773	6.869
<b>6</b>	2.612	2.829	3.143	3.372	3.707	4.317	5.959
<b>7</b>	2.517	2.715	2.998	3.203	3.499	4.029	5.408
<b>8</b>	2.449	2.634	2.896	3.085	3.355	3.833	5.041
<b>9</b>	2.398	2.574	2.821	2.998	3.250	3.690	4.781
<b>10</b>	2.359	2.527	2.764	2.932	3.169	3.581	4.587
<b>11</b>	2.328	2.491	2.718	2.879	3.106	3.497	4.437
<b>12</b>	2.303	2.461	2.681	2.836	3.055	3.428	4.318
<b>13</b>	2.282	2.436	2.650	2.801	3.012	3.372	4.221
<b>14</b>	2.264	2.415	2.624	2.771	2.977	3.326	4.140
<b>15</b>	2.249	2.397	2.602	2.746	2.947	3.286	4.073
<b>16</b>	2.235	2.382	2.583	2.724	2.921	3.252	4.015
<b>17</b>	2.224	2.368	2.567	2.706	2.898	3.222	3.965
<b>18</b>	2.214	2.356	2.552	2.689	2.878	3.197	3.922
<b>19</b>	2.205	2.346	2.539	2.674	2.861	3.174	3.883
<b>20</b>	2.197	2.336	2.528	2.661	2.845	3.153	3.850
<b>21</b>	2.189	2.328	2.518	2.649	2.831	3.135	3.819
<b>22</b>	2.183	2.320	2.508	2.639	2.819	3.119	3.792
<b>23</b>	2.177	2.313	2.500	2.629	2.807	3.104	3.768
<b>24</b>	2.172	2.307	2.492	2.620	2.797	3.091	3.745
<b>25</b>	2.167	2.301	2.485	2.612	2.787	3.078	3.725
<b>26</b>	2.162	2.296	2.479	2.605	2.779	3.067	3.707
<b>27</b>	2.158	2.291	2.473	2.598	2.771	3.057	3.689
<b>28</b>	2.154	2.286	2.467	2.592	2.763	3.047	3.674
<b>29</b>	2.150	2.282	2.462	2.586	2.756	3.038	3.660
<b>30</b>	2.147	2.278	2.457	2.581	2.750	3.030	3.646
<b>40</b>	2.123	2.250	2.423	2.542	2.704	2.971	3.551
<b>60</b>	2.099	2.223	2.390	2.504	2.660	2.915	3.460
<b>120</b>	2.076	2.196	2.358	2.468	2.617	2.860	3.373
$\infty$	2.054	2.170	2.326	2.432	2.576	2.807	3.290



The  $t$ -value with  $\nu = 14$  degrees of freedom that leaves an area of 0.025 to the left, and therefore an area of 0.975 to the right, is

$$t_{0.975} = -t_{0.025} = -2.145$$

### Example

Find  $\Pr(-t_{0.025} < T < t_{0.05})$ .

### Solution

Since  $t_{0.05}$  leaves an area of 0.05 to the right, and  $-t_{0.025}$  leaves an area of 0.025 to the left, we find a total area of  $1 - 0.05 - 0.025 = 0.925$  between  $-t_{0.025}$  and  $t_{0.05}$ .

Hence

$$\Pr(-t_{0.025} < T < t_{0.05}) = 0.925$$

### Example

Find  $k$  such that  $\Pr(k < T < -1.761) = 0.045$  for a random sample of size 15 selected from a normal distribution with

$$T = \frac{\bar{X} - \mu}{S/\sqrt{n}}.$$

### Solution

From Table A.4 we note that 1.761 corresponds to  $t_{0.05}$  when  $\nu = 14$ . Therefore,  $-t_{0.05} = -1.761$ . Since  $k$  in the original probability statement is to the left of  $-t_{0.05} = -1.761$ , let  $k = -t_{\alpha}$ . Then, by using figure, we have

$$0.045 = 0.05 - \alpha, \text{ or } \alpha = 0.005.$$

Hence, from Table A.4 with  $\nu = 14$ ,  
 $k = -t_{0.005} = -2.977$  and  $\Pr(-2.977 < T < -1.761) = 0.045$ .

- 1 Random sampling and statistics
- 2 Sampling Distribution of Means and the Central Limit Theorem
- 3 Sampling Distribution of the Difference between Two Means
- 4 Sampling Distribution of the Variance
- 5 The Student's Distribution
- 6 The Fisher Distribution**
- 7 The Fisher with Two Sample Variances
- 8 Sampling Distribution of Proportions and the Central Limit
- 9 Sampling Distribution of the Difference between Two Proportions

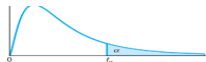
The statistic  $F$  is defined to be the ratio of two independent chi-squared random variables, each divided by its number of degrees of freedom.

**Theorem 31**

The random variable

$$F = \frac{U/\nu_1}{V/\nu_2}$$

where  $U$  and  $V$  are independent random variables having chi-squared distributions with  $\nu_1$  and  $\nu_2$  degrees of freedom, respectively, is the  **$F$ -distribution** with  $\nu_1$  and  $\nu_2$  degrees of freedom (d.f.).

Table A.6 Critical Values of the  $F$ -Distribution

$v_2$	$f_{0.05}(v_1, v_2)$								
	$v_1$								
	1	2	3	4	5	6	7	8	9
1	161.45	199.50	215.71	224.58	230.16	233.99	236.77	238.88	240.54
2	18.51	19.00	19.16	19.25	19.30	19.33	19.35	19.37	19.38
3	10.13	9.55	9.28	9.12	9.01	8.94	8.89	8.85	8.81
4	7.71	6.94	6.59	6.39	6.26	6.16	6.09	6.04	6.00
5	6.61	5.79	5.41	5.19	5.05	4.95	4.88	4.82	4.77
6	5.99	5.14	4.76	4.53	4.39	4.28	4.21	4.15	4.10
7	5.59	4.74	4.35	4.12	3.97	3.87	3.79	3.73	3.68
8	5.32	4.46	4.07	3.84	3.69	3.58	3.50	3.44	3.39
9	5.12	4.26	3.86	3.63	3.48	3.37	3.29	3.23	3.18
10	4.96	4.10	3.71	3.48	3.33	3.22	3.14	3.07	3.02
11	4.84	3.98	3.59	3.36	3.20	3.09	3.01	2.95	2.90
12	4.75	3.89	3.49	3.26	3.11	3.00	2.91	2.85	2.80
13	4.67	3.81	3.41	3.18	3.03	2.92	2.83	2.77	2.71
14	4.60	3.74	3.34	3.11	2.96	2.85	2.76	2.70	2.65
15	4.54	3.68	3.29	3.06	2.90	2.79	2.71	2.64	2.59
16	4.49	3.63	3.24	3.01	2.85	2.74	2.66	2.59	2.54
17	4.45	3.59	3.20	2.96	2.81	2.70	2.61	2.55	2.49
18	4.41	3.55	3.16	2.93	2.77	2.66	2.58	2.51	2.46
19	4.38	3.52	3.13	2.90	2.74	2.63	2.54	2.48	2.42
20	4.35	3.49	3.10	2.87	2.71	2.60	2.51	2.45	2.39
21	4.32	3.47	3.07	2.84	2.68	2.57	2.49	2.42	2.37
22	4.30	3.44	3.05	2.82	2.66	2.55	2.46	2.40	2.34
23	4.28	3.42	3.03	2.80	2.64	2.53	2.44	2.37	2.32
24	4.26	3.40	3.01	2.78	2.62	2.51	2.42	2.36	2.30
25	4.24	3.39	2.99	2.76	2.60	2.49	2.40	2.34	2.28
26	4.23	3.37	2.98	2.74	2.59	2.47	2.39	2.32	2.27
27	4.21	3.35	2.96	2.73	2.57	2.46	2.37	2.31	2.25
28	4.20	3.34	2.95	2.71	2.56	2.45	2.36	2.29	2.24
29	4.18	3.33	2.93	2.70	2.55	2.43	2.35	2.28	2.22
30	4.17	3.32	2.92	2.69	2.53	2.42	2.33	2.27	2.21
40	4.08	3.23	2.84	2.61	2.45	2.34	2.25	2.18	2.12
60	4.00	3.15	2.76	2.53	2.37	2.25	2.17	2.10	2.04
120	3.92	3.07	2.68	2.45	2.29	2.18	2.09	2.02	1.96
$\infty$	3.84	3.00	2.60	2.37	2.21	2.10	2.01	1.94	1.88

Figure : Table A.6 Critical Values of the  $F$ -Distribution

Table A.6 (continued) Critical Values of the  $F$ -Distribution

$v_2$	$f_{0.05}(v_1, v_2)$									
	$v_1$									
	10	12	15	20	24	30	40	60	120	$\infty$
1	241.88	243.91	245.95	248.01	249.05	250.10	251.14	252.20	253.25	254.31
2	19.40	19.41	19.43	19.45	19.45	19.46	19.47	19.48	19.49	19.50
3	8.79	8.74	8.70	8.66	8.64	8.62	8.59	8.57	8.55	8.53
4	5.96	5.91	5.86	5.80	5.77	5.75	5.72	5.69	5.66	5.63
5	4.74	4.68	4.62	4.56	4.53	4.50	4.46	4.43	4.40	4.36
6	4.06	4.00	3.94	3.87	3.84	3.81	3.77	3.74	3.70	3.67
7	3.64	3.57	3.51	3.44	3.41	3.38	3.34	3.30	3.27	3.23
8	3.35	3.28	3.22	3.15	3.12	3.08	3.04	3.01	2.97	2.93
9	3.14	3.07	3.01	2.94	2.90	2.86	2.83	2.79	2.75	2.71
10	2.98	2.91	2.85	2.77	2.74	2.70	2.66	2.62	2.58	2.54
11	2.85	2.79	2.72	2.65	2.61	2.57	2.53	2.49	2.45	2.40
12	2.75	2.69	2.62	2.54	2.51	2.47	2.43	2.38	2.34	2.30
13	2.67	2.60	2.53	2.46	2.42	2.38	2.34	2.30	2.25	2.21
14	2.60	2.53	2.46	2.39	2.35	2.31	2.27	2.22	2.18	2.13
15	2.54	2.48	2.40	2.33	2.29	2.25	2.20	2.16	2.11	2.07
16	2.49	2.42	2.35	2.28	2.24	2.19	2.15	2.11	2.06	2.01
17	2.45	2.38	2.31	2.23	2.19	2.15	2.10	2.06	2.01	1.96
18	2.41	2.34	2.27	2.19	2.15	2.11	2.06	2.02	1.97	1.92
19	2.38	2.31	2.23	2.16	2.11	2.07	2.03	1.98	1.93	1.88
20	2.35	2.28	2.20	2.12	2.08	2.04	1.99	1.95	1.90	1.84
21	2.32	2.25	2.18	2.10	2.05	2.01	1.96	1.92	1.87	1.81
22	2.30	2.23	2.15	2.07	2.03	1.98	1.94	1.89	1.84	1.78
23	2.27	2.20	2.13	2.05	2.01	1.96	1.91	1.86	1.81	1.76
24	2.25	2.18	2.11	2.03	1.98	1.94	1.89	1.84	1.79	1.73
25	2.24	2.16	2.09	2.01	1.96	1.92	1.87	1.82	1.77	1.71
26	2.22	2.15	2.07	1.99	1.95	1.90	1.85	1.80	1.75	1.69
27	2.20	2.13	2.06	1.97	1.93	1.88	1.84	1.79	1.73	1.67
28	2.19	2.12	2.04	1.96	1.91	1.87	1.82	1.77	1.71	1.65
29	2.18	2.10	2.03	1.94	1.90	1.85	1.81	1.75	1.70	1.64
30	2.16	2.09	2.01	1.93	1.89	1.84	1.79	1.74	1.68	1.62
40	2.08	2.00	1.92	1.84	1.79	1.74	1.69	1.64	1.58	1.51
60	1.99	1.92	1.84	1.75	1.70	1.65	1.59	1.53	1.47	1.39
120	1.91	1.83	1.75	1.66	1.61	1.55	1.50	1.43	1.35	1.25
$\infty$	1.83	1.75	1.67	1.57	1.52	1.46	1.39	1.32	1.22	1.00

**Table A.6 (continued) Critical Values of the  $F$ -Distribution**

$v_2$	$f_{0.01}(v_1, v_2)$								
	$v_1$								
	1	2	3	4	5	6	7	8	9
1	4052.18	4999.50	5403.35	5624.58	5763.65	5858.99	5928.36	5981.07	6022.47
2	98.50	99.00	99.17	99.25	99.30	99.33	99.36	99.37	99.39
3	34.12	30.82	29.46	28.71	28.24	27.91	27.67	27.49	27.35
4	21.20	18.00	16.69	15.98	15.52	15.21	14.98	14.80	14.66
5	16.26	13.27	12.06	11.39	10.97	10.67	10.46	10.29	10.16
6	13.75	10.92	9.78	9.15	8.75	8.47	8.26	8.10	7.98
7	12.25	9.55	8.45	7.85	7.46	7.19	6.99	6.84	6.72
8	11.26	8.65	7.59	7.01	6.63	6.37	6.18	6.03	5.91
9	10.56	8.02	6.99	6.42	6.06	5.80	5.61	5.47	5.35
10	10.04	7.56	6.55	5.99	5.64	5.39	5.20	5.06	4.94
11	9.65	7.21	6.22	5.67	5.32	5.07	4.89	4.74	4.63
12	9.33	6.93	5.95	5.41	5.06	4.82	4.64	4.50	4.39
13	9.07	6.70	5.74	5.21	4.86	4.62	4.44	4.30	4.19
14	8.86	6.51	5.56	5.04	4.69	4.46	4.28	4.14	4.03
15	8.68	6.36	5.42	4.89	4.56	4.32	4.14	4.00	3.89
16	8.53	6.23	5.29	4.77	4.44	4.20	4.03	3.89	3.78
17	8.40	6.11	5.18	4.67	4.34	4.10	3.93	3.79	3.68
18	8.29	6.01	5.09	4.58	4.25	4.01	3.84	3.71	3.60
19	8.18	5.93	5.01	4.50	4.17	3.94	3.77	3.63	3.52
20	8.10	5.85	4.94	4.43	4.10	3.87	3.70	3.56	3.46
21	8.02	5.78	4.87	4.37	4.04	3.81	3.64	3.51	3.40
22	7.95	5.72	4.82	4.31	3.99	3.76	3.59	3.45	3.35
23	7.88	5.66	4.76	4.26	3.94	3.71	3.54	3.41	3.30
24	7.82	5.61	4.72	4.22	3.90	3.67	3.50	3.36	3.26
25	7.77	5.57	4.68	4.18	3.85	3.63	3.46	3.32	3.22
26	7.72	5.53	4.64	4.14	3.82	3.59	3.42	3.29	3.18
27	7.68	5.49	4.60	4.11	3.78	3.56	3.39	3.26	3.15
28	7.64	5.45	4.57	4.07	3.75	3.53	3.36	3.23	3.12
29	7.60	5.42	4.54	4.04	3.73	3.50	3.33	3.20	3.09
30	7.56	5.39	4.51	4.02	3.70	3.47	3.30	3.17	3.07
40	7.31	5.18	4.31	3.83	3.51	3.29	3.12	2.99	2.89
60	7.08	4.98	4.13	3.65	3.34	3.12	2.95	2.82	2.72
120	6.85	4.79	3.95	3.48	3.17	2.96	2.79	2.66	2.56
$\infty$	6.63	4.61	3.78	3.32	3.02	2.80	2.64	2.51	2.41



Table A.6 (continued) Critical Values of the  $F$ -Distribution

$v_2$	$f_{0.01}(v_1, v_2)$									
	$v_1$									
	10	12	15	20	24	30	40	60	120	$\infty$
1	6055.85	6106.32	6157.28	6208.73	6234.63	6260.65	6286.78	6313.03	6339.39	6365.86
2	99.40	99.42	99.43	99.45	99.46	99.47	99.47	99.48	99.49	99.50
3	27.23	27.05	26.87	26.69	26.60	26.50	26.41	26.32	26.22	26.13
4	14.55	14.37	14.20	14.02	13.93	13.84	13.75	13.65	13.56	13.46
5	10.05	9.89	9.72	9.55	9.47	9.38	9.29	9.20	9.11	9.02
6	7.87	7.72	7.56	7.40	7.31	7.23	7.14	7.06	6.97	6.88
7	6.62	6.47	6.31	6.16	6.07	5.99	5.91	5.82	5.74	5.65
8	5.81	5.67	5.52	5.36	5.28	5.20	5.12	5.03	4.95	4.86
9	5.26	5.11	4.96	4.81	4.73	4.65	4.57	4.48	4.40	4.31
10	4.85	4.71	4.56	4.41	4.33	4.25	4.17	4.08	4.00	3.91
11	4.54	4.40	4.25	4.10	4.02	3.94	3.86	3.78	3.69	3.60
12	4.30	4.16	4.01	3.86	3.78	3.70	3.62	3.54	3.45	3.36
13	4.10	3.96	3.82	3.66	3.59	3.51	3.43	3.34	3.25	3.17
14	3.94	3.80	3.66	3.51	3.43	3.35	3.27	3.18	3.09	3.00
15	3.80	3.67	3.52	3.37	3.29	3.21	3.13	3.05	2.96	2.87
16	3.69	3.55	3.41	3.26	3.18	3.10	3.02	2.93	2.84	2.75
17	3.59	3.46	3.31	3.16	3.08	3.00	2.92	2.83	2.75	2.65
18	3.51	3.37	3.23	3.08	3.00	2.92	2.84	2.75	2.66	2.57
19	3.43	3.30	3.15	3.00	2.92	2.84	2.76	2.67	2.58	2.49
20	3.37	3.23	3.09	2.94	2.86	2.78	2.69	2.61	2.52	2.42
21	3.31	3.17	3.03	2.88	2.80	2.72	2.64	2.55	2.46	2.36
22	3.26	3.12	2.98	2.83	2.75	2.67	2.58	2.50	2.40	2.31
23	3.21	3.07	2.93	2.78	2.70	2.62	2.54	2.45	2.35	2.26
24	3.17	3.03	2.89	2.74	2.66	2.58	2.49	2.40	2.31	2.21
25	3.13	2.99	2.85	2.70	2.62	2.54	2.45	2.36	2.27	2.17
26	3.09	2.96	2.81	2.66	2.58	2.50	2.42	2.33	2.23	2.13
27	3.06	2.93	2.78	2.63	2.55	2.47	2.38	2.29	2.20	2.10
28	3.03	2.90	2.75	2.60	2.52	2.44	2.35	2.26	2.17	2.06
29	3.00	2.87	2.73	2.57	2.49	2.41	2.33	2.23	2.14	2.03
30	2.98	2.84	2.70	2.55	2.47	2.39	2.30	2.21	2.11	2.01
40	2.80	2.66	2.52	2.37	2.29	2.20	2.11	2.02	1.92	1.80
60	2.63	2.50	2.35	2.20	2.12	2.03	1.94	1.84	1.73	1.60
120	2.47	2.34	2.19	2.03	1.95	1.86	1.76	1.66	1.53	1.38
$\infty$	2.32	2.18	2.04	1.88	1.79	1.70	1.59	1.47	1.32	1.00

### Theorem

Writing  $f_\alpha(\nu_1, \nu_2)$  for  $f_\alpha$  with  $\nu_1$  and  $\nu_2$  degrees of freedom, we have

$$f_{1-\alpha}(\nu_1, \nu_2) = \frac{1}{f_\alpha(\nu_2, \nu_1)}$$

Thus, the  $f$ -value with 6 and 10 degrees of freedom, leaving an area of 0.95 to the right, is  $f_{0.95}(6, 10) = \frac{1}{f_{0.05}(10,6)} = \frac{1}{4.06} = 0.246$ .

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Suppose that random samples of size  $n_1$  and  $n_2$  are selected from two normal populations with variances  $\sigma_1^2$  and  $\sigma_2^2$ , respectively. From Theorem 23, we know that

$$\chi_1^2 = \frac{(n_1 - 1)S_1^2}{\sigma_1^2} \text{ and } \chi_2^2 = \frac{(n_2 - 1)S_2^2}{\sigma_2^2}$$

are random variables having chi-squared distributions with  $\nu_1 = n_1 - 1$  and  $\nu_2 = n_2 - 1$  degrees of freedom. Furthermore, since the samples are selected at random, we are dealing with independent random variables. Then, using Theorem 31 with  $\chi_1^2 = U$  and  $\chi_2^2 = V$ , we obtain the following result.

### Theorem

If  $S_1^2$  and  $S_2^2$  are the variances of independent random samples of size  $n_1$  and  $n_2$  taken from normal populations with variances  $\sigma_1^2$  and  $\sigma_2^2$ , respectively, then

$$F = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2}$$

has an F-distribution with  $\nu_1 = n_1 - 1$  and  $\nu_2 = n_2 - 1$  degrees of freedom.

### Example

For an  $F$ -distribution, find

- (a)  $f_{0.05}$  with  $\nu_1 = 7$  and  $\nu_2 = 15$ ;
- (b)  $f_{0.05}$  with  $\nu_1 = 15$  and  $\nu_2 = 7$ ;
- (c)  $f_{0.01}$  with  $\nu_1 = 24$  and  $\nu_2 = 19$ ;
- (d)  $f_{0.95}$  with  $\nu_1 = 19$  and  $\nu_2 = 24$ ;
- (e)  $f_{0.99}$  with  $\nu_1 = 28$  and  $\nu_2 = 12$ .

### Solution

- (a) 2.71.
- (b) 3.51.
- (c) 2.92.
- (d)  $1/2.11 = 0.47$ .
- (e)  $1/2.90 = 0.34$ .

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In many situations the use of the sample proportion is easier and more reliable because, unlike the mean, the proportion does not depend on the population variance, which is usually an unknown quantity. We will represent the sample proportion by  $\hat{P}$  and the population proportion by  $p$ . Construction of the sampling distribution of the sample proportion is done in a manner similar to that of the mean. One has  $\hat{P} = X/n$  where  $X$  is a number of success for a sample of size  $n$ . It is clear that  $X$  is a binomial distribution  $B(n, p)$ . Its mean  $\mu_X = np$  and its variance  $\sigma_X^2 = np(1 - p)$ .



### Theorem

The mean  $\mu_{\hat{p}}$  of the sample distribution  $\hat{P}$  is equal to the true population proportion  $p$ , and its variance  $\sigma_{\hat{p}}^2$  is equal to  $p(1 - p)/n$ .

### Theorem

If  $np \geq 5$  and  $n(1 - p) \geq 5$ , then the random variable  $\hat{P}$  is approximation a normal distribution with mean  $\mu_{\hat{p}} = p$  and standard deviation (or standard error)  $\sigma_{\hat{p}} = \sqrt{p(1 - p)/n}$ . Hence

$$Z = \frac{\hat{P} - p}{\sqrt{p(1 - p)/n}}$$

is approximately a standard normal distribution.

### Example

In the mid seventies, according to a report by the National Center for Health Statistics, 19.4 percent of the adult U.S. male population was obese. What is the probability that in a simple random sample of size 150 from this population fewer than 15 percent will be obese?

### Solution

Here  $n = 150$ ,  $p = 0.194$ . Since  $np \geq 5$  and  $n(1 - p) \geq 5$ , hence

$$Z = \frac{\hat{P} - 0.194}{\sqrt{0.194(1 - 0.194)/150}} = \frac{\hat{P} - 0.194}{0.032}$$

is approximately a standard normal distribution.

$$\Pr(\hat{P} \leq 0.15) = \Pr\left(\frac{\hat{P} - 0.194}{0.032} \leq \frac{0.15 - 0.194}{0.032}\right) \simeq \Pr(Z \leq -1.37) = 0.0853.$$

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### Theorem

The mean  $\mu_{\hat{p}_1 - \hat{p}_2}$  of the sample distribution of the difference between two sample proportions  $\hat{P}_1 - \hat{P}_2$  is equal to the difference  $p_1 - p_2$  between the true population proportions, and its variance  $\sigma_{\hat{p}_1 - \hat{p}_2}^2$  will be equal to  $p_1(1 - p_1)/n_1 + p_2(1 - p_2)/n_2$ .

### Theorem

If  $n_1 p_1 \geq 5$ ,  $n_1(1 - p_1) \geq 5$ ,  $n_2 p_2 \geq 5$ ,  $n_2(1 - p_2) \geq 5$ , then the random variable  $\hat{P}_1 - \hat{P}_2$  is approximately a normal distribution with mean  $\mu_{\hat{p}_1 - \hat{p}_2} = p_1 - p_2$  and standard deviation (or standard error)  $\sigma_{\hat{p}_1 - \hat{p}_2} = \sqrt{p_1(1 - p_1)/n_1 + p_2(1 - p_2)/n_2}$ . Hence

$$Z = \frac{(\hat{P}_1 - \hat{P}_2) - (p_1 - p_2)}{\sqrt{\frac{p_1(1-p_1)}{n_1} + \frac{p_2(1-p_2)}{n_2}}}$$

is approximately a standard normal distribution.

## Example

Suppose that there are two large high schools, each with more than 2000 students, in a certain town. At School 1, 70% of students did their homework last night. Only 50% of the students at School 2 did their homework last night. The counselor at School 1 takes a sample random sample of 100 students and records the proportion that did homework. School 2's counselor takes a sample random sample of 200 students and records the proportion that did homework. Find the probability of getting a difference in sample proportion  $\hat{P}_1 - \hat{P}_2$  of 0.10 or less from the two surveys.

## Solution

Here  $p_1 = 0.7$ ,  $p_2 = 0.5$ ,  $n_1 = 100$  and  $n_2 = 200$ . It is clear that  $n_1 p_1 \geq 5$ ,  $n_1(1 - p_1) \geq 5$ ,  $n_2 p_2 \geq 5$ ,  $n_2(1 - p_2) \geq 5$ . Also

$$\mu_{\hat{p}_1 - \hat{p}_2} = p_1 - p_2 = 0.2$$

and

$$\sigma_{\hat{p}_1 - \hat{p}_2} = \sqrt{p_1(1 - p_1)/n_1 + p_2(1 - p_2)/n_2} = 0.058$$

. Hence,

$$Z = \frac{\hat{P}_1 - \hat{P}_2 - p_1 - p_2}{\sqrt{\frac{p_1(1-p_1)}{n_1} + \frac{p_2(1-p_2)}{n_2}}} = \frac{\hat{P}_1 - \hat{P}_2 - 0.2}{0.058}$$

is approximately a standard normal.

$$\begin{aligned} \Pr(\hat{P}_1 - \hat{P}_2 \leq 0.10) &= \Pr\left(\frac{\hat{P}_1 - \hat{P}_2 - 0.2}{0.058} \leq \frac{0.10 - 0.2}{0.058}\right) \\ &\simeq \Pr(Z \leq -1.72) = 0.0427. \end{aligned}$$