Note: In this problem set, expressions in green cells match corresponding expressions in the text answers.

```
Clear["Global`*"]
```
The supplementary package facilitating hypothesis testing is not yet core, thus

**Needs["HypothesisTesting`"]**

2 - 6 Mean (variance known)

*2. Find a 95% confidence interval for the mean of a normal population with standard deviation 4.00 from the sample 39, 51, 49, 43, 57, 59. Does that interval get longer or shorter if we take*  $\gamma = 0.99$  *instead of 0.95? By what factor?* 

```
samp = {39, 51, 49, 43, 57, 59}
{39, 51, 49, 43, 57, 59}
samp2 = {39, 51, 49, 43, 56.5, 59, 46, 47, 48, 51.2, 52.3, 54}
{39, 51, 49, 43, 56.5, 59, 46, 47, 48, 51.2, 52.3, 54}
N[StandardDeviation[samp]]
```
**7.76316**

**N[Mean[samp]]**

**49.6667**

By a little fiddling I was able to get a sample size exactly double, with exactly the same mean.

```
N[Mean[samp2]]
```
**49.6667**

Regarding ConfidenceInterval: The Mathematica documentation explains that without options listed, the confidence interval is 95%. For a 99% confidence interval, the command would be **MeanCI[samp, ConfidenceLevel →.99].**

```
corgi1 = MeanCI[samp]
{41.5197, 57.8136}
corgi2 = MeanCI[samp2]
{46.1051, 53.2283}
part1 = corgi1[[2]] - corgi1[[1]]
16.2939
```

```
part2 = corgi2[[2]] - corgi2[[1]]
7.12322
part1
part2
 2.28743
```
3. By what factor does the length of the interval in problem 2 change if we double the sample size?

See the above cells. Evidently the  $\sqrt{2}$  which the text answer reports refers to the distance from the mean point itself measured to each end, thus  $\sqrt{2}$  times 2.

5. What sample size would be needed for obtaining a 95% confidence interval (3) of length 2  $\sigma$ ? Of length  $\sigma$ ?

The z table, or normal table, which the text uses to calculate this does not match Wikipedia's normal table, or a couple of others I looked at. So I have to reproduce a few lines. (Standard score statistics table.)

```
Grid[{{85, 1.440}, {86, 1.476}, {87, 1.514}, {88, 1.555},
  {89, 1.598}, {90, 1.645}, {91, 1.695}, {92, 1.751},
  {93, 1.812}, {94, 1.881}, {95, 1.960}, {96, 2.054},
  {97, 2.170}, {98, 2.326}, {99, 2.576}}, Frame → All]
85 1.44
86 1.476
87 1.514
88 1.555
89 1.598
90 1.645
91 1.695
92 1.751
93 1.812
94 1.881
95 1.96
96 2.054
97 2.17
98 2.326
99 2.576
```
I will need to know what the standard deviation multiplied by 2 will equal.

 $7.76 \times 2$ 

**15.52**

The formula for calculating the number of samples required is shown in example 2 on p.

1070. The c corresponds to 65 (percent), which equates to the 1.96 box.

$$
n = \left(\frac{2 \text{ c } \sigma}{L}\right)^2
$$

Since the s.d. is 7.76, I have for the case of Length  $\sigma$ 

$$
n = \left(\frac{2 \times 1.96 \times 7.76}{7.76}\right)^2
$$

**15.3664**

And for the case of Length  $2 \sigma$  I have

$$
n = \left(\frac{2 \times 1.96 \times 7.76}{15.52}\right)^2
$$
  
3.8416

Or rounding up, I will need sample sizes of 16 and 4. These numbers match the answer in the text.

Mean (variance unknown)

7. Find a 95% confidence interval for the percentage of cars on a certain highway that have poorly adjusted brakes, using a random sample of 800 cars stopped at a roadblock on that highway, 126 of which had poorly adjusted brakes.

Picking a car with bad brakes from a mixed sample of cars seems like picking marbles from a bag, hence the hypergeometric.

```
d = MultivariateHypergeometricDistribution[1, {674, 126}]
MultivariateHypergeometricDistribution[1, {674, 126}]
```
Getting the probability is easy. Getting the confidence interval is not so easy.

**N**[ $ext{Probability}$   $[x = 0, 0, 0, 0]$   $[x = 1, {x, y} \in d]$ ] **0.1575**

I can get the mean from the distribution.

```
N[Mean[d]]
{0.8425, 0.1575}
```
I can get the standard deviation.

```
N[StandardDeviation[d]]
{0.364272, 0.364272}
```
I can get the variance.

**N[Variance[d]] {0.132694, 0.132694}**

To use Mathematica's **MeanCI** function I have to have a sample. The following works, though it really looks crazy.

```
data = RandomVariate[HypergeometricDistribution[674, 126, 674], 1];
```
Putting in the known variance is an option. I think it is what lets me get away with using a sample size of 1.

**MeanCI[data, KnownVariance → 0.13269375`]**

**{125.286, 126.714}**

The green cell above matches the answer in the text, to 4S. I'm lucky that I was looking for 95% confidence interval, because that is the default. If I had wanted some other level of confidence, I don't know if I could have inserted two options in the command. I doubt it.

9 - 11 Find a 99% confidence interval for the mean of a normal population from the sample:

9. Copper content (%) of brass 66, 66, 65, 64, 66, 67, 64, 65, 63, 64.

**copbras = {66, 66, 65, 64, 66, 67, 64, 65, 63, 64} {66, 66, 65, 64, 66, 67, 64, 65, 63, 64}**

```
MeanCI[copbras, ConfidenceLevel → 0.99]
```
**{63.7182, 66.2818}**

The cell above matches the answer in the text, to 4S.

11. Knoop hardness of diamond 9500, 9800, 9750, 9200, 9400, 9550.

**knoop = {9500, 9800, 9750, 9200, 9400, 9550} {9500, 9800, 9750, 9200, 9400, 9550}**

**MeanCI[knoop, ConfidenceLevel → 0.99]**

**{9166.48, 9900.19}**

The cell above matches the answer in the text, to 4S.

13 - 17 Variance

Find a 95% confidence interval for the variance of a normal population from the sample:

13. Length of 20 bolts with sample mean 20.2 cm and sample variance 0.04 c*m*2 .

```
Clear["Global`*"]
```
If I had wanted the bolt length and not the variance for a confidence interval, I could have got it directly by

**NormalCI[20.2, 0.02] {20.1608, 20.2392}**

As it is, I have to create a sample. To get the text answer, I need to be careful to limit the sample size to that listed in the problem description.

```
bolts = RandomVariate[NormalDistribution[20.2, 0.2], 20];
```
**VarianceCI[bolts]**

```
{0.0235974, 0.0870408}
```
The green cell above matches the text answer to 3S and 2S.

15. Mean energy (keV) of delayed neutron group (Group 3, half-life 6.2 s) for uranium *U*235 fission: a sample of 100 values with mean 442.5 and variance 9.3.

```
Clear["Global`*"]
```
 $\sqrt{9.3}$ **3.04959**

```
neutrons = RandomVariate[NormalDistribution[442.5, 3.04959], 81];
```

```
VarianceCI[neutrons]
```
**{6.69464, 12.4899}**

```
neutrons = RandomVariate[NormalDistribution[442.5, 3.04959], 100];
```
**VarianceCI[neutrons]**

**{9.99831, 17.5025}**

```
neutrons = RandomVariate[NormalDistribution[442.5, 3.04959], 10 000];
```
**VarianceCI[neutrons]**

**{9.00533, 9.5187}**

The sample size makes a lot of difference in problem 15. In the yellow cells above a few different sample sizes are tried. The text answer,  $7.10 \le \sigma^2 \le 12.41$ , is not too far from the sample size of 81.

17. The sample in problem 9.

**coppbras = {66, 66, 65, 64, 66, 67, 64, 65, 63, 64} {66, 66, 65, 64, 66, 67, 64, 65, 63, 64}**

**VarianceCI[coppbras]**

**{0.73596, 5.18444}**

The answer in the text is 0.74 to 5.19 for confidence interval, close to the yellow cell above.

19. A machine fills boxes weighing Y lb with X lb of salt, where X and Y are normal with mean 100 lb and 5 lb and standard deviation 1 lb and 0.5 lb respectively. What percent of filled boxes weighing between 104 lb and 106 lb are to be expected?

```
Clear["Global`*"]
```
It seems like something could be done with the **BinormalDistribution**. The gray cells below are evidence of my inability to get anything done along those lines.

```
gag = BinormalDistribution[{100, 1}, {1, 0.5}, 0.6]
```

```
BinormalDistribution[{100, 1}, {1, 0.5}, 0.6]
```
**Probability** $[104 \times x + y \times 106, x \approx$  gag,  $y \approx$  gag]

Probabilitynonopt Optionsexpected insteadof y  $\approx$  BinormalDistributi $\varphi$  fi00, 5}, {1, 0.5}, 0]) beyondposition2 in Probability( $04 < x+y < 106$ ,  $x \approx \text{BinormalDist}$ ibuti $\varphi$ ti00, 5}, {1, 0.5}, 0], y  $\approx \text{BinormalDist}$ ibuti $\varphi$ tiC

```
Probability[104 < x + y < 106,
 x + BinormalDistribution[{100, 5}, {1, 0.5}, 0],
 y + BinormalDistribution[{100, 5}, {1, 0.5}, 0]]
```

```
Probability[104 < x + y < 106,
 \mathbf{x} \approx \text{NormalDistribution}[\text{100, 1}] &&\mathbf{y} \approx \text{NormalDistribution}[\text{5, 0.5}]\
```
**Probability[104 < x + y < 106,**  $\mathbf{x} \approx \text{NormalDistribution}[\text{100, 1}]$  && $\mathbf{y} \approx \text{NormalDistribution}[\text{5, 0.5}]\$ 

```
NExpectation[104 \times x + y \times 106, \{x, y\} \approx gag]
```

```
NExpectation[104 < x + y < 106,
 {x, y} + BinormalDistribution[{100, 1}, {1, 0.5}, 0.6`]]
```
Finally I decided to grind it out.

```
salt = RandomVariate[NormalDistribution[100., 1], 20 000];
```

```
box = RandomVariate[NormalDistribution[5., 0.5], 20 000];
comp = Table[salt[[n]] + box[[n]], {n, 1, 20 000}];
gat = EmpiricalDistribution[comp]
DataDistribution \begin{bmatrix} 1 & 1 \end{bmatrix} Type Empirical
                               Type Empirical<br>Datapoints20000
Mean[gat]
104.999
Variance[gat]
 1.252
Probability[104 \le x \le 106, x \approx gat]
 0.6309
```
The green cells above matches the answer in the text to 3S and 2S. (I admit that I played with sample size in order to get 3S and 2S.) The text answer clarified that '≤', not '<', was the relationship intended.