

Uncertainties02B.wmxm

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1 *Preface*

In Uncertainties02B.wmxm we introduce further concepts in error analysis, with an emphasis on the physical sciences.

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2 *References*

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Frederick James, *'Statistical Methods in Experimental Physics'*, 2nd ed.,
 2006, World Scientific.

Louis Lyons, *Statistics for Nuclear and Particle Physics*, 1986, Cambridge Univ. Press,

Luca Lista, *'Statistical Methods for Data Analysis in Particle Physics'*,
Lecture Notes in Physics 909, 2016, Springer-Verlag,

Glen Cowan , *Statistical Data Analysis*, Clarendon Press, Oxford, 1998

```
(%i6) load(draw)$
set_draw_defaults(line_width=2, grid = [2,2], point_type = filled_circle,
  head_type = 'nofilled, head_angle = 20, head_length = 0.5,
  background_color = light_gray, draw_realpart=false)$
load (descriptive)$ load (distrib)$
fpprintprec : 5$ ratprint : false$
```

Homemade functions fill, head, tail, Lsum are useful for looking at long lists.

```
(%i10) fill ( aL ) := [ first (aL), last (aL), length (aL) ]$
head(L) := if listp (L) then rest (L, - (length (L) - 3) ) else
  error("Input to 'head' must be a list of expressions ")$
tail (L) := if listp (L) then rest (L, length (L) - 3 ) else
  error("Input to 'tail' must be a list of expressions ")$
Lsum (aList) := apply ("+", aList)$
```

3 *fracData (List, a, b)*

fracData (alist, a, b) calculates the fraction of the list numbers which lie in the interval [a, b].

```
(%i11) fracData (myData, xx1, xx2) :=
  block ([ ccnt : 0 ],
    for j thru length (myData) do
      if myData[j] >= xx1 and myData[j] <= xx2 then ccnt : ccnt + 1,
    float (ccnt/ length (myData)) )$
```

4 *fracDataLessThan (data, maxval)*

```
(%i12) fracDataLessThan (myData,maxval) :=
  block ([ ccnt : 0 ],
    for j thru length (myData) do
      if myData[j] <= maxval then ccnt : ccnt + 1,
    float (ccnt/ length (myData)) )$
```

5 *sigE (List)*

sigE (data) returns the standard deviation of the mean (SDOM), σE of a data set.

```
(%i13) sigE(data) := float ( std1 (data) / sqrt (length (data)) )$
```

6 *confidence (q, m, s)*

With q a number in the interval $0 < q < 1$, and with m the mean and s the standard deviation of a Normal distribution, confidence (q, m, s) prints out the values $dx, m - dx, m + dx$, and outputs a list $[m - dx, m + dx]$ which allows one to have $100 \cdot q$ % confidence a random value of x will lie within that interval, ie., within $m \pm dx$.

```
(%i14) confidence (qq, mm, ss) :=
  block ([ddx],
    ddx :
      float ( (quantile_normal (qq + (1 - qq)/2, mm, ss) -
        quantile_normal ( (1 - qq)/2, mm, ss))/2 ),
    print ( "delx = ", ddx, " = ", sconcat (ddx/ss, " *  $\sigma$ " ) ),
    print ( " x1 = ", mm - ddx, ", x2 = ", mm + ddx ),
    [mm - ddx, mm + ddx] )$
```

7 *The Standard Error of the Mean σE , SDOM*

We repeat the definition of the standard error of the mean presented in the previous worksheet, based on [HH] Sec. 2.7. The 'standard error of the mean' is also called the 'statistical error on the mean'; see Lyons, Sec. 1.5. Fred Senese, Sec. 8.3.1, uses the symbol sE for the standard error of the mean. [HH] use the symbol δ (Greek delta).

"The width of the histogram of means is a measure of the precision of the mean. It is clearly evident from the above plots that the width of the histogram of the means decreases as the size of the sample used to calculate the mean increases—this is a consequence of averaging over statistical fluctuations."

"The width of the histogram of means is the standard deviation of the mean, also known as the standard error, σE ."

"When the number of measurements involved in calculating the mean increases, the means are better defined; consequently the histograms of the distributions of the means are narrower. Note that the precision with which the mean can be determined is related to the number of measurements used to calculate the mean."

"In practice one does not generate histograms of the means based on trials containing many measurements; rather one uses all N measurements x_i to calculate one value for the mean $\langle x \rangle$. We expect the standard error (the standard deviation of the mean), σE , to decrease as the number of data points we collect, N , increases."

It can be shown "that the standard error is reduced by a factor of \sqrt{N} with respect to the sample standard deviation:"

$$\sigma E = \sigma_{\text{sample}} / N^{1/2}$$

"A data set containing N multiple readings yields one value of the mean. Thus we should quote our findings as the mean \pm the error on the mean, i.e."

$$\langle x \rangle \pm \sigma E == \langle x \rangle \pm \sigma_{\text{sample}} / N^{1/2}$$

"In other words, we are saying that there is a two-thirds chance that the measured parameter is within the range [$\langle x \rangle - \sigma E$, $\langle x \rangle + \sigma E$]."

"One can interpret the standard error as being a standard deviation, not of the measurements, but rather the mean: this is why the standard error σE is also called the standard deviation of the mean (SDOM)."

8 *The Error in the Error*

[HH] Sec. 2.7.1

"There exists a formula for the fractional error in the error (Squires 2001, Appendix B); it is defined as

$$(\text{fractional}) \text{ error in the error} = 1 / \sqrt{2*N - 2},$$

and is a slowly decreasing function of N."

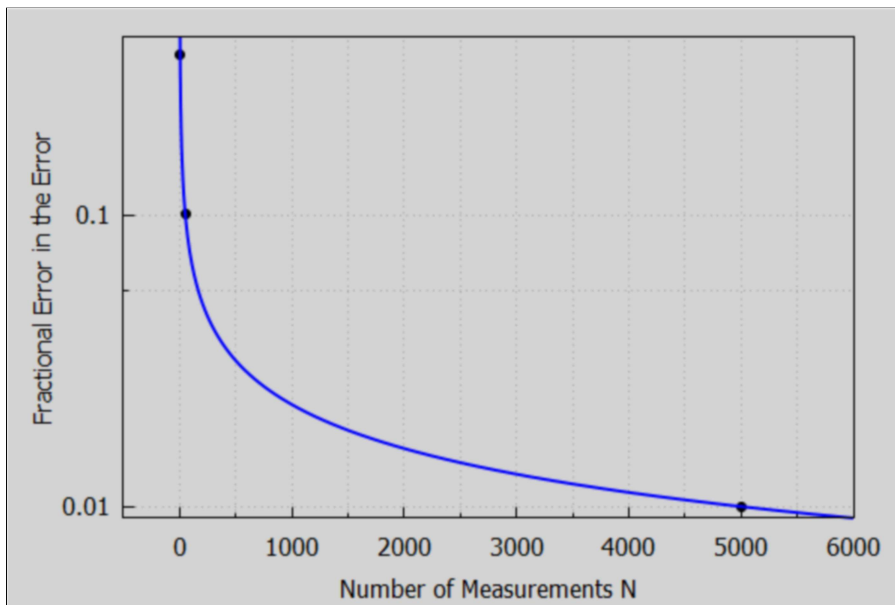
```
(%i15) mypts : makelist ( [N, float ( 1 / sqrt ( 2*N - 2 ) ) ], N, [5, 50, 5000] );
```

```
(mypts) [[5,0.35355],[50,0.10102],[5000,0.010001]]
```

With N = 5 points, the fractional error in the error estimate is 0.35, with N = 50 points the fractional error in the error estimate is 0.1, with N = 5000 points, 0.01.

```
(%i16) wxdraw2d ( xrange = [-500, 6000], logy = true, color = black,
  xlabel = "Number of Measurements N",
  ylabel = "Fractional Error in the Error",
  points (mypts), color = blue, explicit (1/sqrt(2*N-2), N, 4, 6e3) )$
```

```
(%t16)
```



"It should be noted that the error on the error is a slowly decreasing function with respect to N. For example, with only five measurements the error estimate is only good to about 1 part in 3 (35%). As the sample size N increases, the error in the error decreases, and we can be more confident in our results, allowing for more significant figures to be quoted. Note that the error in the error does not fall to a few percent (allowing two significant figures to be quoted meaningfully) until approximately 10,000 data points have been collected (see Exercise 2.4)."

"Conversely, care should be taken when choosing the number of appropriate significant figures [of the error] if the first significant figure of the error is 1 — rounding an error of 1.4 to 1, or 1.51 to 2 causes a change in the error of approximately 25%. "

"The following rule is generally adopted:

Quote the error to one significant figure.

Corollary:

- (i) If you have collected approximately 10,000 data points, or more, consider quoting the error to two significant figures;
- (ii) if the first significant figure of the error is 1, consider quoting the second significant figure."

"Worked example:

Analysis of repeated measurements of the acceleration due to gravity, g , yields $\langle g \rangle = 9.812\,3456\text{ m s}^{-2}$, with $\sigma_E = 0.032\,1987\text{ m s}^{-2}$.

** If this answer was based on 10 measurements, you would report

$$g = (9.81 \pm 0.03)\text{ m s}^{-2}.$$

** If this answer was based on 7,500 measurements you would consider reporting

$$g = (9.812 \pm 0.032)\text{ m s}^{-2}.$$

If another measurement technique has results which are

$$\langle g \rangle = 9.817\,654\text{ m s}^{-2}\text{ with } \sigma_E = 0.101\,23\text{ m s}^{-2},\text{ then}$$

** If this answer was based on 10 measurements, you would report

$$g = (9.8 \pm 0.1)\text{ m s}^{-2};$$

** if this answer was based on 500 measurements, you would consider reporting

$$g = (9.82 \pm 0.10)\text{ m s}^{-2}."$$

"Note that it is extremely rare to quote errors to three significant figures or higher. As we saw earlier, even the currently accepted values for the fundamental constants have their errors quoted only to two significant figures. Note also that there is no rule about how many significant figures are included in the mean—this is ascertained after the error (and its error) are evaluated. The value of the acceleration due to gravity deduced in the worked example after 7500 measurements has an error known to two significant figures, and a mean known to four significant figures, whereas Avogadro's number has an error known to two significant figures, and a mean known to nine significant figures."

"From the preceding section, we can formulate the following procedures to be considered when we quote our results:

- (1) Analyze the experimental data and calculate the mean; keep all significant figures at this stage.
- (2) Calculate the standard error σ_E (the error in the mean); keep all significant figures at this stage.
- (3) Think about how many significant figures should be retained for the error σ_E having reflected on the number of data points collected.
- (4) Round the mean to the appropriate decimal place."

8.1 Ten Measurements of the Period of a Pendulum

[HH] Sec. 2.3.1: Measurement of the period of a pendulum expressed in seconds.
"Rough-and-ready approach to estimating the width"

Ten measurements of the period of a pendulum (in sec.) were made, with the raw data being:

```
(%i17) trial : cons ("Trial", makelist (j,j,1,10) );
```

```
(trial) [ Trial, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10 ]
```

```
(%i18) data : [10.0, 9.4, 9.8, 9.6, 10.5, 9.8, 10.3, 10.2, 10.4, 9.3];
```

```
(data) [ 10.0, 9.4, 9.8, 9.6, 10.5, 9.8, 10.3, 10.2, 10.4, 9.3 ]
```

```
(%i19) matrix (trial, cons ("Period", data) );
```

```
(%o19) 
$$\begin{pmatrix} \textit{Trial} & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ \textit{Period} & 10.0 & 9.4 & 9.8 & 9.6 & 10.5 & 9.8 & 10.3 & 10.2 & 10.4 & 9.3 \end{pmatrix}$$

```

```
(%i20) Lsum (data) / length (data);
```

```
(%o20) 9.93
```

"Rough-and-ready approach to estimating the width"

"All data points lie within the interval $9.3 \leq T \leq 10.5$ s which covers a range of 1.2 s, or a spread around the mean ($\langle T \rangle = 9.9$ s) of about ± 0.6 s. Evaluating the maximum spread of the data is one rough-and-ready approach to estimating the precision of the measurement. It is somewhat pessimistic, because as we will see later, we generally take the precision in a measurement to be two-thirds of the maximum spread of values, and therefore the spread of the data is approximately ± 0.4 s. The factor of two thirds is justified for a Gaussian distribution. Note that the spread of the measurements is significantly worse than the precision of the measuring instrument—this is why taking repeat measurements is important."

"We can therefore say that:

- * a typical measurement of the period is likely to be within 0.4 seconds of the mean value;
- * the precision of the measurements is 0.4 seconds."

With the Maxima tools we calculate the mean, sample standard deviation, and the standard error of the mean σ_E .

```
(%i22) m : mean (data);
       $\sigma_{\text{sample}}$  : std1 (data);
```

```
(m) 9.93
```

```
( $\sigma_{\text{sample}}$ ) 0.41913
```

```
(%i23)  $\sigma_E$  :  $\sigma_{\text{sample}}$  / sqrt (length (data) ), numer;
```

```
( $\sigma_E$ ) 0.13254
```

```
(%i24) sigE (data);
```

```
(%o24) 0.13254
```

"As the analysis is based on so few data points only the first significant figure of the standard error of the mean is retained; the result is reported as $T = (9.9 \pm 0.1)$ s"

8.2 HH Exer. 2.4

Recall the advice given above about choosing the number of significant figures of the standard error σE to report.

"Conversely, care should be taken when choosing the number of appropriate significant figures [of the error] if the first significant figure of the error is 1 — rounding an error of 1.4 to 1, or 1.51 to 2 causes a change in the error of approximately 25%. "

"The following rule is generally adopted:

Quote the error to one significant figure.

Corollary:

- (i) If you have collected approximately 10,000 data points, or more, consider quoting the error to two significant figures;
- (ii) if the first significant figure of the error is 1, consider quoting the second significant figure."

"Consider a set of measurements with the standard error calculated to be $\sigma E = 0.987\ 654\ 321$. Here we address the question of how many significant figures should be quoted. Construct a spreadsheet with four columns. The first column should be N , the number of measurements on which σE is based. In the second column write σE to the nine significant figures quoted above. The third column should be $\sigma E * (1 - 1/\sqrt{2*N - 2})$, and the fourth column should be $\sigma E * (1 + 1/\sqrt{2*N - 2})$."

"As we are interested in the variation over a large dynamic range, choose values for N such as 2, 3, 5, 10, 20, 30, etc. Verify the statement above that the number of data points, N , needs to approach a few tens of thousands before the second significant figure in the error can be quoted, i.e. when the values in the three columns become equal to the second significant figure. REPEAT the analysis for the case where $\sigma E = 0.123\ 456\ 789$, i.e. the first significant digit of the error is 1. How many data points must be collected before the third significant figure can be quoted?"

In the first case the first significant digit of the standard error is 9.

```
(%i26)  $\sigma E : 0.987654321$ $
      f (N) := print (N,  $\sigma E$ ,  $\sigma E * \text{float} ( (1 - 1/\sqrt{2*N - 2}) )$ ,
                     $\sigma E * \text{float} ((1 + 1/\sqrt{2*N - 2}) )$  )$
```

```
(%i27) fpprintprec : 2$
```

```
(%i28) for n in [10,20,100, 200,1000,5000,10000,20000,30000,50000,
70000] do f(n)$
10 0.99 0.75 1.2
20 0.99 0.83 1.1
100 0.99 0.92 1.1
200 0.99 0.94 1.0
1000 0.99 0.97 1.0
5000 0.99 0.98 1.0
10000 0.99 0.98 0.99
20000 0.99 0.98 0.99
30000 0.99 0.98 0.99
50000 0.99 0.98 0.99
70000 0.99 0.99 0.99
```

For $N = 70,000$ data points and when restricting the printed values to two significant figures, we finally have the three columns having the same value, and we can report the standard error of the mean as (after rounding) 0.99.

REPEAT the analysis for the case where $\sigma E = 0.123\ 456\ 789$, i.e. the first significant digit of the error is 1. How many data points must be collected before the third significant figure can be quoted?"

```
(%i29)  $\sigma E : 0.123456789$ $
```

```
(%i30) fpprintprec : 3$
```

```
(%i31) for n in [10,20,100, 200,1000,5000,10000,20000] do f(n)$
10 0.123 0.0944 0.153
20 0.123 0.103 0.143
100 0.123 0.115 0.132
200 0.123 0.117 0.13
1000 0.123 0.121 0.126
5000 0.123 0.122 0.125
10000 0.123 0.123 0.124
20000 0.123 0.123 0.124
```

We see that with 10,000 data points, the three columns agree and we can quote $\sigma E = 0.123$ to three significant figures.

9 *Rounding and Significant Figures*

[HH] Sec. 2.8.1

"RULE OF THUMB: If an error is not quoted assume that the uncertainty is in the last reported digit."

"The theme of this chapter has been that all measurements are subject to uncertainty. A working rule is that, in the absence of an error being quoted, we assume that a number has significance equal to a single unit in the last figure quoted."

"Thus if we were to say that the resistance of a resistor was 97Ω , it is said to have an absolute uncertainty of 1Ω and is said to be known to two significant figures. A resistor with a value of 100.04Ω indicates an absolute uncertainty of 0.01Ω and is said to be known to five significant figures. Confusion can occur in ascertaining how many significant figures a number has when zeroes are involved."

"Rules for identifying significant digits"

- " * All non-zero digits are significant:
 $2.998 \times 10^8 \text{ m s}^{-1}$ has four significant figures.
- * All zeroes between non-zero digits are significant:
 $6.022\ 141\ 79 \times 10^{23} \text{ mol}^{-1}$ has nine significant figures.
- * Zeroes to the left of the first non-zero digits are not significant:
 0.51 MeV has two significant figures.
- * Zeroes at the end of a number to the right of the decimal point are significant:
 $1.60 \times 10^{-19} \text{ C}$ has three significant figures.
- * If a number ends in zeroes without a decimal point, the zeroes might be significant:
 270 might have two or three significant figures."

RULE OF THUMB: To avoid confusion when numbers end in zeros, report your values using scientific notation.

"The ambiguity in the last rule to use if a number ends in zeroes can be resolved by the use of scientific notation. For example, depending on whether two or three significant figures is appropriate, we could write 270Ω as $0.27 \text{ k}\Omega$, or $2.7 \times 10^2 \Omega$, both of which have two significant figures; or $0.270 \text{ k}\Omega$, or $2.70 \times 10^2 \Omega$, both of which have three significant figures. Note that the entries $0.3 \text{ k}\Omega$ and 300Ω in a lab book carry very different significance."

"Significant figures must also be considered when carrying out calculations. It is important to carry all digits through to the final result before rounding to avoid rounding errors which compromise the accuracy of the final result. The principle is the following: the precision of a calculated result is limited by the least precise measurement in the calculation."

9.1 Rounding Rules

"Decide which is the last digit to keep, then:

- ** Leave the last digit unchanged if the next digit is 4 or lower: 6.62×10^4 becomes 6.6×10^4 if only two significant figures are appropriate.
- ** Increase the last digit by 1 if the next digit is 6 or higher: 5.67×10^8 becomes 5.7×10^8 if only two significant figures are appropriate.

If the digit after the last one to be retained is 5 the recommended procedure is to choose the even round value:

- ** Leave the last digit unchanged if it is even. For example: 3.45 becomes 3.4 if only two significant figures are appropriate.
- ** Increase the last digit by 1 if it is odd. For example: 3.55 becomes 3.6 if only two significant figures are appropriate.

This 'round-to-even method' avoids bias in rounding, because half of the time we round up, and half of the time we round down."

"In addition and subtraction, the result is rounded off to the same number of decimal places as the number with the least number of decimal places. For example, $1.23 + 45.6$ should be quoted as 46.8. This reflects the fact that we do not know whether the 45.6 is 45.56 or 45.64 to the next decimal place."

"In multiplication and division, the answer should be given to the same number of significant figures as the component with the least number of significant figures. For example, 1.2×345.6 is evaluated as 414.72 but quoted as 4.1×10^2 on account of the least precise value having only two significant figures."

"It is important to carry all significant figures through long calculations to avoid unnecessary rounding errors. Rounding to the appropriate precision should only be done at the end of the calculation."

10 *Uncertainties as Probabilities*

We follow [HH], Ch. 3.

[HH] Sec 3.2.2

"A box contains 100 Ω resistors which are said to have a standard deviation of 2 Ω .

- 1.) What is the probability of selecting a resistor with a value of 95 Ω or less?
- 2.) What is the probability of finding a resistor in the range 99 to 101 Ω ?"

We can assume the box contains resistors which are certified by the factory to have $\langle R \rangle = 100 \Omega$ and $\sigma E = 2 \Omega$ = standard deviation of the mean. Since the value of the resistance R can take on a continuum of values, the Gaussian (Normal) distribution is the appropriate distribution for error analysis. Since there is no information of the size of the sample used by the manufacturer to determine the SDOM σE , we cannot tell what the sample standard deviation $\sigma_{\text{sample}} = \text{std1}(\text{sample})$ was.

As we know from Stat04-Normal.wmx, we can use the Maxima function `pdf_normal (x, <x>, σ)`. The Maxima function `pdf_normal (x, μ , σ)` returns a symbolic value for the value of the Normal (μ , σ) probability distribution at position x . μ is the mean = $\langle x \rangle = E(x) = 100 \Omega$, σ is the standard deviation of the mean, which we take here as $\sigma E = \sigma_{\text{sample}}/\sqrt{N}$, in which $\sigma_{\text{sample}} = \sqrt{(N - 1)^{-1} \sum (x_i - \langle x \rangle)^2}$, which is computed using the Maxima function `std1` for a given data set. In particular, here σE is given as 2Ω .

As was discussed in Stat04-Normal.wmx,

Roughly 68% of the area under the Normal (μ , σ) curve lies in the region $[\mu - \sigma, \mu + \sigma]$, ie., $\mu - \sigma \leq x \leq \mu + \sigma$.

Roughly 95% of the area under the Normal (μ , σ) curve lies in the region $[\mu - 2\sigma, \mu + 2\sigma]$, ie., $(\mu - 2\sigma) \leq x \leq (\mu + 2\sigma)$.

Roughly 99.7% of the area under the Normal (μ , σ) curve lies in the region $[\mu - 3\sigma, \mu + 3\sigma]$, ie., $(\mu - 3\sigma) \leq x \leq (\mu + 3\sigma)$.

The probability of selecting a resistor with a value of 95Ω or less is the integral of `pdf_normal (x, 100, 2)` from $x = -\infty$ to $x = 95$, for which we can use the 'cumulative distribution function' `cdf_normal (xfinal, <x>, σ)`.

`cdf_normal (x1, μ , σ)` gives the probability that the value of x measured lies in the interval $[-\infty, x1]$, given that the values measured are described by the normal distribution with mean μ and standard deviation σ .

```
(%i32) cdf_normal (95, 100, 2), numer;
```

```
(%o32) 0.00621
```

`cdf_normal` returns a probability which is always a real number greater than or equal to zero and less than or equal to one. $0 \leq p \leq 1$. Multiplying the probability by 100 gives the chance or likelihood of finding that a randomly chosen resistor will have a resistance of less than or equal to 95Ω . Here that chance is about 0.6%.

The longer path to this answer is to directly integrate `pdf_normal`:

```
(%i33) integrate (pdf_normal (x, 100, 2), x, minf, 95);
```

```
(%o33) 
$$\frac{\sqrt{2}\sqrt{\pi} - \sqrt{2}\sqrt{\pi} \operatorname{erf}\left(\frac{5}{2^{3/2}}\right)}{2^{3/2}\sqrt{\pi}}$$

```

```
(%i34) float (%);
```

```
(%o34) 0.00621
```

The probability of finding a randomly chosen resistor in the range 99 to 101 Ω can be found by summing (integrating) over the probabilities $P(x) dx$ of finding x in $[x, x+dx]$ starting at $x = 99$, ending at $x = 101$, with x taking a continuum of floating point values.

```
(%i35) float ( integrate (pdf_normal (x, 100, 2), x, 99, 101) );
```

```
(%o35) 0.383
```

or we could use `cdf_normal` (why does this work?):

```
(%i36) cdf_normal (101, 100, 2) - cdf_normal (99, 100, 2), numer;
```

```
(%o36) 0.383
```

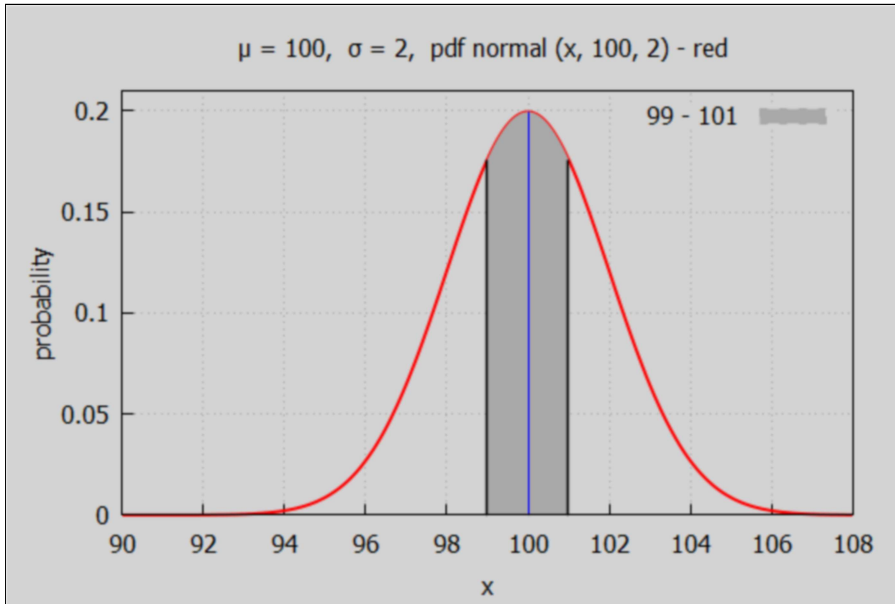
The chance of finding a randomly chosen resistor from the box to have a resistance in the range [99 Ω , 101 Ω] is about 38%.

The area under the curve of Normal (100, 2) with x in the range [99, 101] gives the probability 0.383. We can show this using `wxdraw2d`.

```
(%i37) pdf_normal (100, 100, 2), numer;
```

```
(%o37) 0.199
```

```
(%i38) wxdraw2d (xrange = [90, 108], yrange = [0, 0.21],
  xlabel = "x", ylabel = "probability", grid = true,
  title = " μ = 100, σ = 2, pdf normal (x, 100, 2) - red ",
  color = red, line_width = 2, explicit (pdf_normal (x, 100, 2), x, 90, 108),
  filled_func = true, fill_color = dark_gray, key = " 99 - 101 ",
  explicit ( pdf_normal (x, 100, 2), x, 99, 101),
  color = black, line_width = 1.5, key = "",
  parametric (99,yy,yy,0, pdf_normal(99, 100, 2) ),
  parametric (101,yy,yy,0, pdf_normal(101, 100, 2) ), line_width = 1,
  color = blue, parametric (100,yy,yy,0, pdf_normal(100, 100, 2) ) )$
```



(%t38)

$3\sigma = 3 \cdot 2 = 6$. Notice for $x < 92$ and $x > 106$ there is negligible area (probability) under the curve.

11 Confidence Intervals

The Maxima function `confidence (q, m, s)` is defined at the top of this worksheet. With q a number in the interval $0 < q < 1$, and with m the mean and s the standard deviation of a Normal distribution, `confidence (q, m, s)` prints out the values dx , $m - dx$, $m + dx$, and outputs a list `[m - dx, m + dx]` which allows one to have $100 \cdot q$ % confidence a random value of x will lie within that interval, ie., within $m \pm dx$.

```
(%i39) fpprintprec;
```

```
(%o39) 3
```

Approximately 68% confidence that x will be found in range $100 \pm 1\sigma$;
 " we are confident, at the 68% level, that, were we to take another measurement, the value would lie within one standard deviation of the mean."

(%i40) confidence (0.68, 100, 2)\$
 $delx = 1.99 = 0.994 * \sigma$
 $x1 = 98.0, x2 = 1.02 \cdot 10^2$

(%i41) confidence (0.683, 100, 2)\$
 $delx = 2.0 = 1.0 * \sigma$
 $x1 = 98.0, x2 = 1.02 \cdot 10^2$

About 90% confidence that x will be found in range $100 \pm 1.65\sigma$

(%i42) confidence (0.90, 100, 2)\$
 $delx = 3.29 = 1.64 * \sigma$
 $x1 = 96.7, x2 = 1.03 \cdot 10^2$

(%i43) confidence (0.902, 100, 2)\$
 $delx = 3.31 = 1.65 * \sigma$
 $x1 = 96.7, x2 = 1.03 \cdot 10^2$

(%i44) confidence (0.95, 100, 2)\$
 $delx = 3.92 = 1.96 * \sigma$
 $x1 = 96.1, x2 = 1.04 \cdot 10^2$

1.96 approx = 2.0, so approximately 95% confidence that x will be found in the range $100 \pm 2\sigma$.

(%i45) confidence (0.955, 100, 2)\$
 $delx = 4.01 = 2.0 * \sigma$
 $x1 = 96.0, x2 = 1.04 \cdot 10^2$

(%i46) confidence (0.99, 100, 2)\$
 $delx = 5.15 = 2.58 * \sigma$
 $x1 = 94.8, x2 = 1.05 \cdot 10^2$

(%i47) confidence (0.997, 100, 2)\$
 $delx = 5.94 = 2.97 * \sigma$
 $x1 = 94.1, x2 = 1.06 \cdot 10^2$

(%i48) confidence (0.9973, 100, 2)\$
 $delx = 6.0 = 3.0 * \sigma$
 $x1 = 94.0, x2 = 1.06 \cdot 10^2$

Approximate Confidence Summary Table

(%i49) **matrix** (
 ["Centered on mean", "+/- σ", "+/- 1.65 σ", "+/- 2σ", "+/- 2.58 σ", "+/- 3σ"],
 ["Measurements within range", "68%", "90%", "95%", "99%", "99.7%"],
 ["Measurements outside range", "32%", "10%", "5%", "1.0%", "0.3%"],
 [" ", "1 in 3", "1 in 10", "1 in 20", "1 in 100", "1 in 400"]);

(%o49)
$$\left(\begin{array}{cccccc} \textit{Centered on mean} & +/- \sigma & +/- 1.65 \sigma & +/- 2\sigma & +/- 2.58 \sigma & +/- 3\sigma \\ \textit{Measurements within range} & 68\% & 90\% & 95\% & 99\% & 99.7\% \\ \textit{Measurements outside range} & 32\% & 10\% & 5\% & 1.0\% & 0.3\% \\ & 1 \textit{ in } 3 & 1 \textit{ in } 10 & 1 \textit{ in } 20 & 1 \textit{ in } 100 & 1 \textit{ in } 400 \end{array} \right)$$

"Recalling the interpretation of the standard error as a standard deviation of the mean we can also calculate other confidence intervals. Whereas often in the physical sciences the error bar is taken as one standard deviation of the mean (the standard error), other conventions exist; in other disciplines the 95% confidence limit is often used."

We have shown above "that 95.0% of the measurements lie within the range $\pm 1.96\sigma$. Therefore if a data set of N measurements has a mean $\langle x \rangle$ and standard deviation $\sigma_{\text{sample}} == \sigma_{(N-1)}$, we would report the result at the 95% confidence limit as"
 $\langle x \rangle +/- 1.96 * \sigma E == \langle x \rangle +/- 1.96 * \sigma_{(N-1)} / \sqrt{N}$

"When $\sigma_{(N-1)}$ is ascertained from experimental data, especially from a small number of repeat measurements, greater care is needed with confidence limits." Later we will discuss the 'Student's t distribution', which is more appropriate for confidence interval estimation from a small number of data points."

11.1 random_normal (μ, σ), random_normal (μ, σ, n)

The Maxima function random_normal (m, s) returns a Normal (m, s) random variate. The parameter m is the requested average (mean) value of the distribution. The parameter s is the requested standard deviation of the distribution.

The Maxima function random_normal (m, s, n) returns a list of n random Normal (m, s) variates, where m is the average (mean) of the distribution and s is one standard deviation.

For example, here is a list of 10 normally distributed floating point (real) numbers taken from a Gaussian distribution whose mean is 100 and standard deviation is 2.

```
(%i51) fpprintprec : 3$
      rsample : random_normal (100, 2, 10);

(rsample) [98.8,99.7,1.01 102,99.6,1.0 102,1.01 102,100.0,99.2,98.9,97.5]
```

We simulate a sample consisting of 1,000 resistors randomly chosen. We avoid printing out the entire list of 1,000 values by ending the command with the dollar sign (\$). We can then use our homemade functions (defined at the top of the worksheet) fill, head, and tail.

```
(%i55) rsample : random_normal (100, 2, 1000)$
      fill (rsample);
      head (rsample);
      tail (rsample);

(%o53) [1.02 102,98.8,1000]
(%o54) [1.02 102,1.0 102,99.2]
(%o55) [99.7,98.3,98.8]
```

Each time you run this command, you will (in general) get a different set of 1,000 resistor values. For this particular run we calculate the data list values. Of course the number of values will always be 1,000, but the mean and the value of one standard deviation will in general be different.

```
(%i58) num : length (rsample);
      m : mean (rsample);
      s : std1 (rsample);

(num) 1000
(m) 100.0
(s) 2.05
```

Our Maxima function fracData (alist, a, b), defined at the top of this worksheet, calculates the fraction of the list numbers which lie in the interval [a, b].

For this sample to be faithful to a Gaussian (normal) distribution, about 68% of the data values should be in the range $[\langle x \rangle - \sigma, \langle x \rangle + \sigma] = [98, 102]$.

The fraction of the sample within $\pm \sigma$ of the mean (theor: 0.68):

```
(%i59) fracData (rsample, 98, 102);
(%o59) 0.663
```

The sample fraction in range 99 ohms - 101 ohms (theory: 0.38).

```
(%i60) fracData (rsample, 99, 101);
(%o60) 0.381
```

The sample fraction with $R \leq 95$ ohms (theory: 0.006).

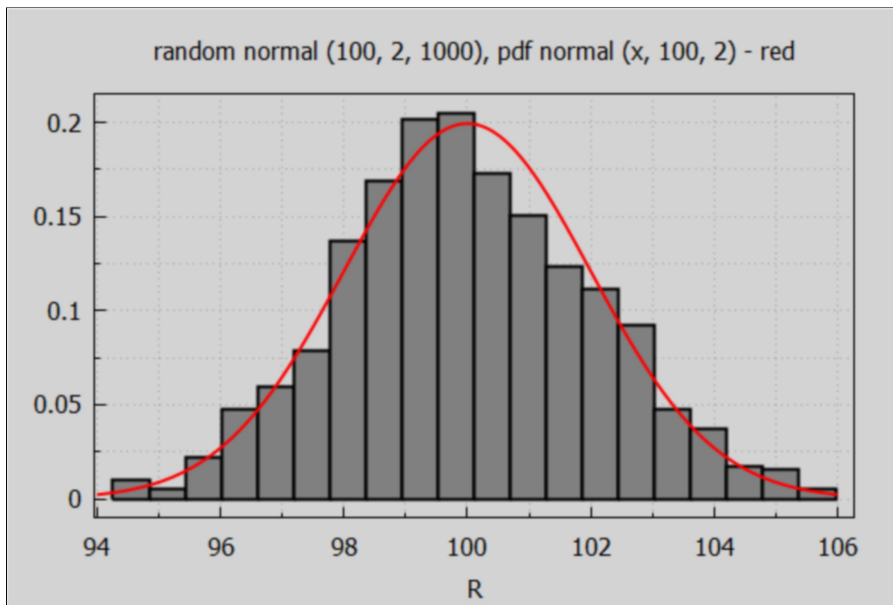
```
(%i61) fracDataLessThan (rsample, 95);
(%o61) 0.007
```

The comparisons between the sample and theory would be closer if we used a simulation using 10,000 or 100,000 random_normal values.

As we did in Stat04-Normal.wmx, we plot a histogram using rsample, choosing to throw the raw data into 20 bins. We show a plot of pdf_normal (x, 100, 2) in red also.

```
(%i62) wxdraw2d( xrange = [94, 106], xlabel = "R",
  title = " random normal (100, 2, 1000), pdf normal (x, 100, 2) - red ",
  histogram_description (rsample, nclasses = 20, frequency = density,
    fill_color = black, fill_density = 0.5), color = red,
  explicit(pdf_normal (x, 100, 2), x, 94, 106) )$
```

(%t62)



12 Comparisons: Experiment vs. Accepted Value

Reference: [HH] 3.3.4

- 1.) If your experimental result and the accepted value differ by up to one standard error, they are 'in excellent agreement.'
- 2.) If your experimental result and the accepted value differ by between one and two standard errors, they are 'in reasonable agreement.'
- 3.) If your experimental result and the accepted value differ by more than three standard errors, they are 'in disagreement.'