

Phys 238: Uncertainties resulting from Random, Independent errors

```
In[72]:= Clear["Global`*"]
```

Individual Random Numbers

```
In[73]:= RandomReal[{5, 7}] (* Generates a random number between 5 and 7. *)
```

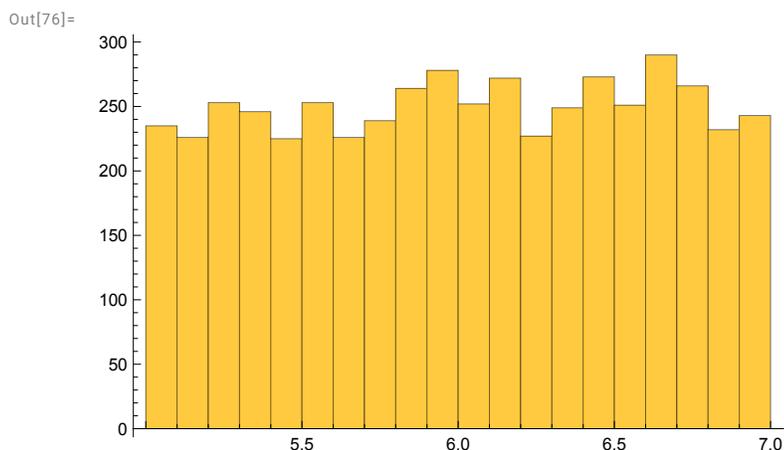
```
Out[73]=  
5.20891
```

```
In[74]:= RandomReal[{5, 7}, 10] (* Generate a bunch (10) of them *)
```

```
Out[74]=  
{5.06997, 5.86994, 5.60286, 5.67317,  
5.31659, 6.96606, 5.6477, 6.87101, 5.27136, 5.4901}
```

```
In[75]:= vals = RandomReal[{5, 7}, 5000]; (* A big bunch *)
```

```
In[76]:= Histogram[vals]
```



Repeating a single measurement that results from the sum of many individual random numbers should give a normal distribution.

Define a function that represents one data point that is a result of many independent random numbers. In this example, “many” = 200.

```
In[77]:= onepoint := Mean[RandomReal[{5, 7}, 200]]
```

This returns a different measurement each time you try it.

```
In[78]:= onepoint
```

```
Out[78]=  
6.0662
```

```
In[79]:= onepoint
```

```
Out[79]=
```

```
6.01492
```

Now imagine doing 5000 such measurements.

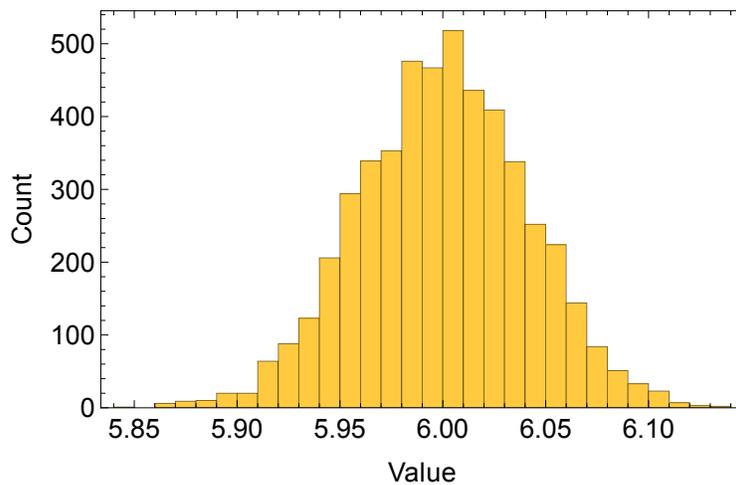
```
In[80]:= data = Table[onepoint, {i, 1, 5000}];
```

Display a histogram with useful plot labels, and in a larger size (scaled to 60% of the screen size here). Give this graph a name so we can use it again later.

```
In[128]:=
```

```
histplot = Histogram[data, Frame → True,
  FrameLabel → {"Value", "Count"}, LabelStyle → Larger, ImageSize → Scaled[0.6]]
```

```
Out[128]=
```



```
In[82]:=  $\mu$  = Mean[data]
```

```
 $\sigma$  = StandardDeviation[data]
```

```
 $\delta\mu$  =  $\sigma$  / Sqrt[Length[data]]
```

```
Out[82]=
```

```
5.99995
```

```
Out[83]=
```

```
0.0408734
```

```
Out[84]=
```

```
0.000578037
```

How many are within $\pm 1\sigma$ of the mean? Use the 'Select' function to select data points within σ of μ .

```
In[85]:= nin1 = Length[Select[data,  $\mu - \sigma \leq \# \leq \mu + \sigma$  &]] / Length[data];
```

```
StringForm["`", or "`%", nin1, NumberForm[100.0 * nin1, 5]]
```

```
Out[86]=
```

```
 $\frac{3387}{5000}$ , or 67.74%
```

Now try to model this distribution by a Gaussian. Getting a manual list of the histogram numbers is a bit tedious. The HistogramList[] function actually gives two separate lists, first is a list of bin boundaries, and second is the list of counts in each bin.

```
In[87]:= bins, counts = N[HistogramList[data]]
Out[87]=
{{5.84, 5.85, 5.86, 5.87, 5.88, 5.89, 5.9, 5.91, 5.92,
  5.93, 5.94, 5.95, 5.96, 5.97, 5.98, 5.99, 6., 6.01, 6.02, 6.03,
  6.04, 6.05, 6.06, 6.07, 6.08, 6.09, 6.1, 6.11, 6.12, 6.13, 6.14},
{1., 0., 6., 9., 10., 20., 20., 64., 88., 123., 206., 294., 339., 353., 476.,
  467., 518., 436., 409., 338., 252., 224., 144., 84., 51., 33., 23., 7., 3., 2.}}
```

```
In[88]:= bins
Out[88]=
{5.84, 5.85, 5.86, 5.87, 5.88, 5.89, 5.9, 5.91, 5.92,
  5.93, 5.94, 5.95, 5.96, 5.97, 5.98, 5.99, 6., 6.01, 6.02, 6.03,
  6.04, 6.05, 6.06, 6.07, 6.08, 6.09, 6.1, 6.11, 6.12, 6.13, 6.14}
```

```
In[89]:= counts
Out[89]=
{1., 0., 6., 9., 10., 20., 20., 64., 88., 123., 206., 294., 339., 353., 476.,
  467., 518., 436., 409., 338., 252., 224., 144., 84., 51., 33., 23., 7., 3., 2.}
```

Because the 'bins' array specifies the boundaries, the length of the 'bins' array is actually one more than the length of the 'counts' array.

```
In[90]:= Length[bins]
Length[counts]
Out[90]=
31
Out[91]=
30
```

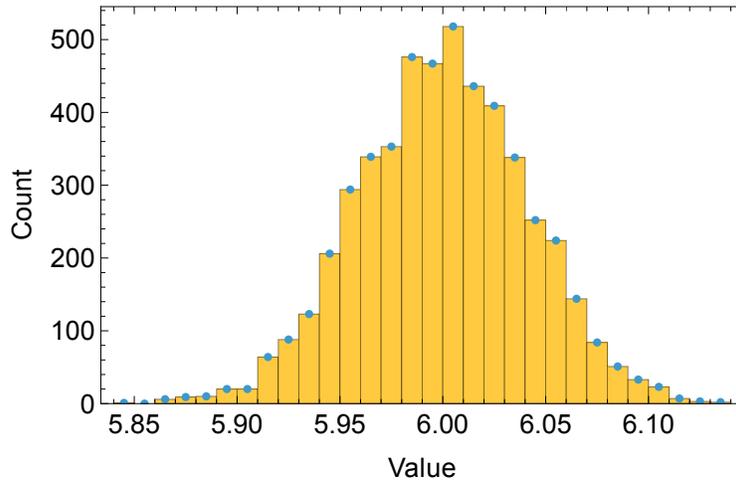
Interleaving them is straightforward, if verbose. We will assign each height to the center point of each bin by constructing a new table.

```
In[92]:= hist = Table[{(bins[[i]] + bins[[i + 1]]) / 2., counts[[i]],
{i, 1, Length[counts]}}];
```

In[129]:=

```
(* Show those data points superimposed on our histogram
to make sure everything looks right. Note that the Show[ ]
command takes a list of plots (in curly braces) to display. *)
Show[{histplot, ListPlot[hist]}
```

Out[129]=



Enter an unnormalized gaussian (Eq. 5.20 in Taylor's text):

In[115]:=

```
gauss[A_, μ_, σ_, x_] :=  $\frac{A}{\sigma \sqrt{2\pi}} \text{Exp}\left[-\frac{(x - \mu)^2}{2\sigma^2}\right]$ 
```

In[131]:=

```
Clear[A, μ, σ]
(* Fit the data to our gaussian *)
result = NonlinearModelFit[hist, gauss[A, μ, σ, x],
  {A, μ, σ}, x]
```

⋯ **NonlinearModelFit:** Failed to converge to the requested accuracy or precision within 100 iterations.

Out[132]=

```
FittedModel[  $3.66 \times 10^3 e^{-0.0496 (-\langle\langle 19 \rangle\rangle + x)^2}$  ]
```

In[133]:=

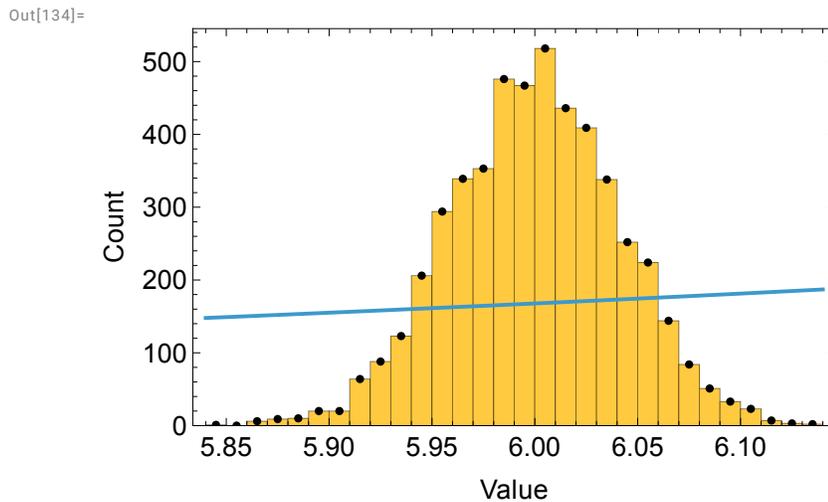
```
result["BestFitParameters"]
```

Out[133]=

```
{A → 29 090.7, μ → 13.882, σ → 3.17514}
```

Show the histogram, the data used in the fit, and the best-fit gaussian all on the same graph. Add useful titles. This fit is very poor.

```
In[134]:=
randerrplot = Show[{
  histplot,
  ListPlot[hist, PlotStyle -> Black], Plot[result[x], {x, Min[bins], Max[bins]}]
}]
]
```



The initial fit attempt does always work well. It is useful to seed the process with guesses for the ' μ ' and ' σ ' parameters. These don't have to be exact, but it helps if they are close. Here, we can just use the calculated mean and standard deviation.

```
In[135]:=
Clear[A,  $\mu$ ,  $\sigma$ ]
result = NonlinearModelFit[hist, gauss[A,  $\mu$ ,  $\sigma$ , x],
  {A,
   { $\mu$ , Mean[data]},
   { $\sigma$ , StandardDeviation[data]}
  }, x]
```

Out[136]=

FittedModel[$486. e^{-296. (\ll 1 \gg)^2}$]

```
In[137]:=
result["BestFitParameters"]
```

Out[137]=

{A -> 50.0578, μ -> 6.00027, σ -> 0.0411088}

```
In[138]:=
result["ParameterConfidenceIntervalTable"]
```

Out[138]=

	Estimate	Standard Error	Confidence Interval
A	50.0578	0.76618	{48.4857, 51.6298 }
μ	6.00027	0.000726524	{5.99878, 6.00176 }
σ	0.0411088	0.000726595	{0.039618, 0.0425997 }

These are very close to the calculated mean and standard deviation:

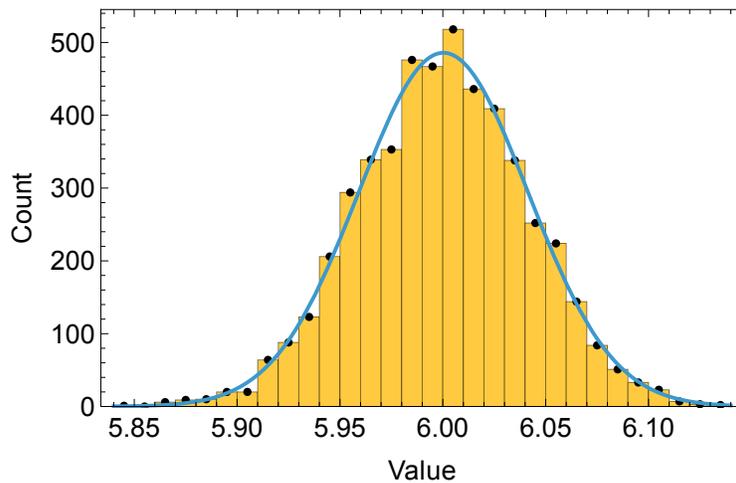
```
In[139]:= {Mean[data], StandardDeviation[data]}
```

```
Out[139]= {5.99995, 0.0408734}
```

Show the histogram, the data used in the fit, and the best-fit gaussian all on the same graph. Add useful titles.

```
In[140]:= randerrplot = Show[{  
  histplot,  
  ListPlot[hist, PlotStyle → Black], Plot[result[x], {x, Min[bins], Max[bins]}]  
}]
```

```
Out[140]=
```



```
In[141]:= SetDirectory[NotebookDirectory[]] (* Save it in this directory *)
```

```
Out[141]= /Users/doughera/238/2026/lectures-dev/Ch05
```

```
In[142]:= Export["randerrplot.pdf", randerrplot]
```

```
Out[142]= randerrplot.pdf
```