

the Hurst Exponent ... and financial stuff

motivated by e-mail from Carl P.

Once upon a time, a British government bureaucrat named Harold Edwin Hurst studied 800 years of records of the Nile's flooding. He noticed that there was a tendency for a high flood year to be followed by another high flood year, and for a low flood year to be followed by another low flood year.

Was that accidental ... or was there *really* some correlation between levels?
Did the height at year 5 have an effect on the height in year 6?

>Are you talking about river levels ... or financial stuff?
Patience.

To analyze, we might do something like this:

1. Note the heights of the n flood levels:
h(1), **h**(2), ... **h**(n)

2. Let **m** be the **M**ean of these levels:
M = (1/n) [**h**(1)+**h**(2)+...+**h**(n)]
Calculate the deviations from the mean:
x(1) = **h**(1) - **M**
x(2) = **h**(2) - **M**
...
x(n) = **h**(n) - **M**
Note that the set of **x**s have zero mean.
Positive **x**'s indicate that the Nile level was above the average.

3. Now calculate the Sums:
Y(1) = **x**(1)
Y(2) = **x**(1) + **x**(2)
...
Y(n) = **x**(1) + **x**(2) + ...+ **x**(n)
Note that the set of partial sums, the **Y**'s, are sums of zero-mean variables.
They will be positive if there's a preponderance of positive **x**'s.
Note, too, that **Y**(k) = **Y**(k-1) + **x**(k).

4. Let **R**(n) = MAX[**Y**(k)] - MIN[**Y**(k)]
This difference between the maximum and minimum of the n values is called the **Range**

5. Let **s**(n) be the standard deviation of the set of n **h**-values.

As it turns out, the probability theorist [William Feller](#) proved that if a series of random variables (like the **x**'s) had finite standard deviation and were independent, then the so-called **R/s** statistic (formed over n observations) would increase in proportion to **n**^{1/2} (for large values of n).

>Huh? The so-called R/s statistic?
Yes. Apparently lots of people are interested in this animal. (See this [PDF stuff](#))
This guy, **R/s**, is called the [rescaled range](#)

Anyway, we now have:
R(n) / **s**(n) ~ k**n**^{1/2} ... where k is some constant

If that were true, then we'd expect that:

log(**R/s**) ~ log(k) + (1/2) log(**n**)

So, if we were to plot log(**R/s**) vs log(**n**), we'd expect it to be approximately a straight line with slope (1/2).

>A logarithm to what base?
It doesn't matter.

Anyway, what Hurst apparently found, was that the plot had a slope closer to 0.7 (rather than 0.5).

>So, what's that mean?
I guess it means that the annual Nile levels weren't independent, but this year's level might be expected to affect next year's level.
Indeed, if the slope of the $\log(\mathbf{R} / \mathbf{s})$ vs $\log(\mathbf{n})$ "best fit line" is **H**, then we'd expect:

$\mathbf{R} / \mathbf{s} \sim \mathbf{k}\mathbf{n}^{\mathbf{H}}$

>Don't tell me! That **H** is the Hurst Exponent, right?
You got it.

>So what's it got to do with financial stuff?
Patience.

>So where's the spreadsheet?
Patience.

The interesting thing is that many things seem to exhibit this long term patterns or dependence ... seven years of plenty followed by seven years of plenty.

>Sounds like a biblical reference.
Yes. It's called the **Joseph Effect**

>Do you realize that don't have a single picture? A picture is worth a thousand ...

Hurst Examples

Okay, let's look at 300 daily returns for Exxon stock.
We'll call them $h(1), h(2), \dots, h(300)$.
We calculate the Mean of these 300 returns. We'll call it **M**.
 $M = (1/300) [h(1) + h(2) + \dots + h(300)]$
Then we calculate $x(1), x(2), \dots, x(300)$, the 300 deviations from the Mean:
 $x(1) = h(1) - M, x(2) = h(2) - M, \dots, x(300) = h(300) - M$.
These deviations are shown in **green**, in Figure 1.
(The average of these deviations is zero!)

Now we calculate the Y's::
 $Y(1) = x(1), Y(2) = x(1)+x(2), \dots, Y(300) = x(1)+x(2)+\dots+x(300)$.
The Y's are shown in **red**, in Figure 1.

Now we find the maximum Y and the minimum Y and subtract them.
That's the Range, $\mathbf{R} = \text{Max}[Y] - \text{Min}[Y]$... in **blue**.
Finally we calculate the Standard Deviation of the h's:
 $s = \text{STDEV}[h(k)]$

>And you get a Hurst exponent ... somehow?

Okay, from the above scheme we note two magic numbers:
 $\mathbf{n} = 300, \mathbf{R} / \mathbf{s} = 0.225 / 0.0462 = 4.87$.

That'll give us one point on our $\log(\mathbf{R}/\mathbf{s})$ vs $\log(\mathbf{n})$ chart, namely
 $\log(300) = \mathbf{5.70}$ and $\log(4.87) = \mathbf{1.58}$.

Now we repeat the above scheme for 310 points, then 320 points *etc. etc.*, each time generating a point on our chart, and ...

>Just give us the chart, okay?
Okay, see Figure 2 →
You see our first point? We calculated points up to $n = 550$.

>And the Hurst exponent is ... uh, the slope?
Yes. At least it's an *estimate* of the Hurst Exponent **H** = 0.478.

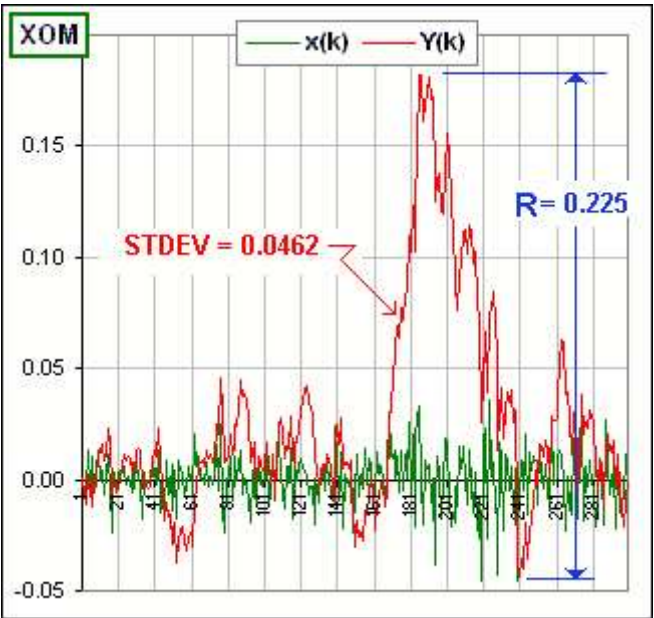


Figure 1

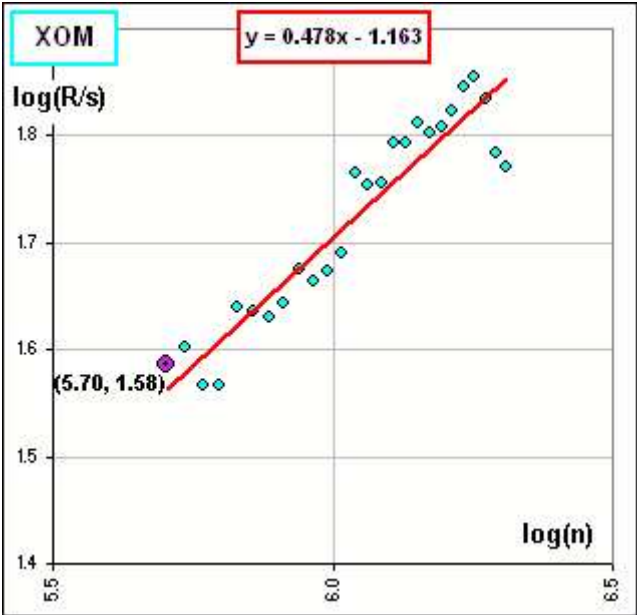


Figure 2

>Pretty close to 1/2, eh?

Yes, and that'd imply that daily returns for XOM are random, uncorrelated, a Brownian motion, independent ...

>Yeah, yeah. Do you always get that?

Patience ...



for Part II

For a great read on Hurst stuff, see bearcave.com

gummy stuff