Trendlines and Numerical Issues in Google Spreadsheet

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Abstract

Google Spreadsheet is an online spreadsheet program in the style of Microsoft Excel. It provides much of Excel's data management and charting functionality, though not all, and also allows spreadsheets to be shared and edited cooperatively. Perhaps the most powerful feature of Google Spreadsheets is the API, which allows the owner of a spreadsheet to programatically modify and read it. Using that API, Google Spreadsheet can provide a turnkey and user friendly window into arbitrary data as it is gathered. The number of Google apps users is growing, with at least 40 million users[1] that have apps accounts, and some of those users will depend on Spreadsheet's accuracy for their business. This paper will examine Google Spreadsheet's accuracy at computing linear best fit lines with ill-conditioned data, comparing it with GNU Octave and Antony Kaplan's work on Excel [2] as well as some related quirks of Google Spreadsheet.

[1] Google Apps for Enterprise: http://www.google.com/enterprise/apps/business/
[2] Kaplan 2010 http://cs.nyu.edu/courses/spring12/CSCI-UA.0421-001/kaplanproject.pdf

Floating Point Numbers in Google Spreadsheet

It can be difficult to tell what sort of floating point math a spreadsheet program uses. Microsoft Excel, for instance, uses cosmetic rounding to make binary floating point numbers appear to be decimal floating point numbers [3]. Google, on the other hand, made no attempt to confirm or deny the floating-point truth. To the best of my knowledge, exhibited below, Google uses software implemented decimal floating point.

If we look at the edge of Google Spreadsheet's precision in binary, we see that $1 + 2^{-46} = 1 + 2^{-47} \neq 1$. This suggests that Google's floating point is not in the form $m * 2^x$.

The edge of Google Spreadsheet's precision with binary floating point:

i	2^{-i}	$1 + 2^{-i}$	$1 = 1 + 2^{-i}$	$1 + 2^{-i} = 1 + 2^{-(i+1)}$
44	0.000000000000057	1.00000000000006	FALSE	FALSE
45	0.000000000000028	1.00000000000003	FALSE	FALSE
46	0.000000000000014	1.00000000000001	FALSE	TRUE
47	0.000000000000007	1.00000000000001	FALSE	FALSE
48	0.000000000000004	1	TRUE	TRUE
49	0.000000000000002	1	TRUE	TRUE
50	0.000000000000001	1	TRUE	TRUE
51	0	1	TRUE	TRUE

In decimal we see no such irregularities:

The edge of Google Spreadsheet's precision with decimal floating point:

i	10^{-i}	$1 + 10^{-i}$	$1 = 1 + 10^{-i}$	$1 + 10^{-i} = 1 + 10^{-(i+1)}$
12	0.000000000001	1.000000000001	FALSE	FALSE
13	0.0000000000001	1.0000000000001	FALSE	FALSE
14	0.00000000000001	1.00000000000001	FALSE	FALSE
15	0.000000000000001	1	TRUE	TRUE

[3] How Futile are Mindless..., Errors designed not to be found: http://www.cs.berkeley.edu/ wkahan/Mindless.pdf

This gives us a few identities:

$$1 + 2^{-47} \neq 1$$
$$1 + 2^{-48} = 1$$
$$1 + 2^{-46} = 1 + 2^{-47} = 1 + 10^{-14}$$
$$1 = 1 + 10^{-15}$$

Where Excel used hardware binary floating point and hushed it up, Google seems to have implemented true decimal floating point.

The TREND function

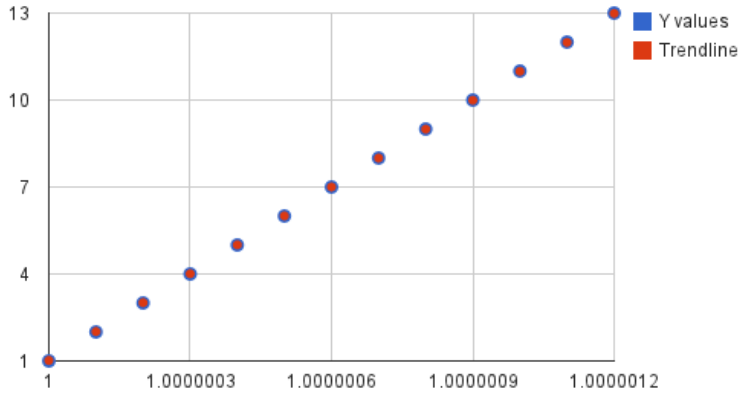
Google Spreadsheet's basic linear best fit function is TREND(data_X,data_Y,x). Given two vectors of data, it calculates the value of the linear best fit at x. However, as the data it is given becomes ill conditioned, the accuracy of TREND desintigrates. Here we examine data series' of the form $(1 + i/10^{-j}, i + 1)$, where i ranges from 0 to 12.

Data series for j=6

X Values	Y values	Trendline
1	1	1
1.00000001	2	1.93142622709274
1.00000002	3	2.86285246908665
1.00000003	4	3.79427869617939
1.00000004	5	4.72570492327213
1.00000005	6	5.65713113546372
1.00000006	7	6.58855739235878
1.00000007	8	7.51998360455036
1.00000008	9	8.4514098316431
1.00000009	10	9.38283605873585
1.0000001	11	10.3142623007298
1.00000011	12	11.2456885278225
1.00000012	13	12.1771147549152

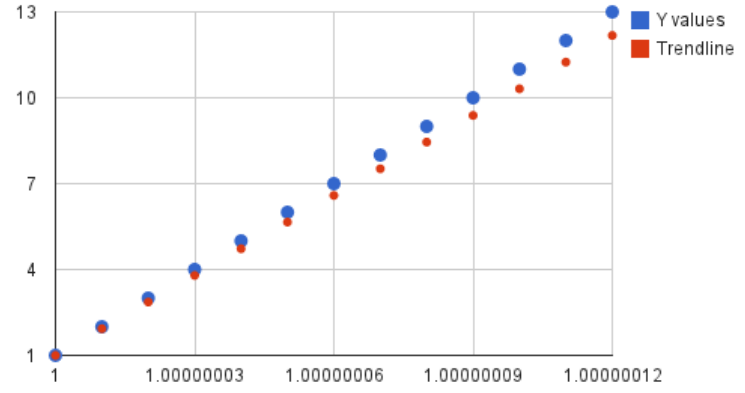
When $j = 0$, there is no numerical error ($R^2 = 1$). However, by the time $j = 5$, we begin to see error creep into the numbers ($R^2 = 0.999999999999965$). By the time $j = 7$, only one correct decimal place remains, creating barely visible error ($R^2 = 0.99999438640566$).

Figure 1: $(1 + i/10^7, i + 1)$ in Google Spreadsheet



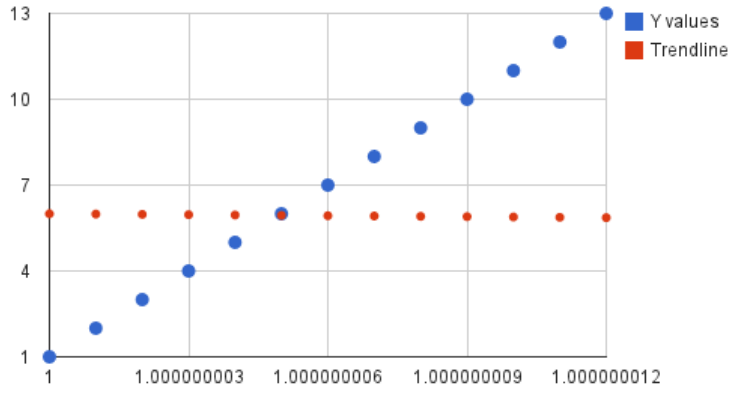
When $j = 8$, the trendline has ventured out alone, with an R^2 value of 0.983604355317896.

Figure 2: $(1 + i/10^{-8}, i + 1)$ in Google Spreadsheet



At $j = 9$, there is only madness. Rather than a slope of 10^j, the trendline has a small negative slope, and $R^2 = -0.078210891033562$.

Figure 3: $(1 + i/10^9, i + 1)$ in Google Spreadsheet

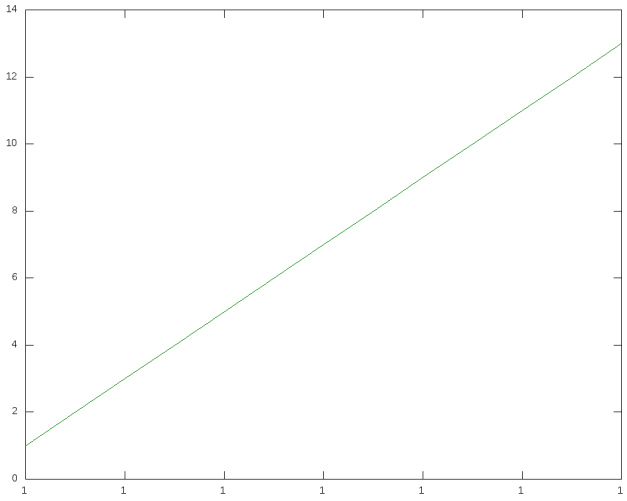


At $j = 10$, and any $j > 10$ which I tried, the spreadsheet crashes with the not-particularly-enlightening message: "Oops. A server error occurred. The page will reload.", and the page reloads, with the change forgotten. This is a consistent result.

By contrast, GNU Octave performs perfectly [4] up to $j = 14$, the edge of its numerical precision.

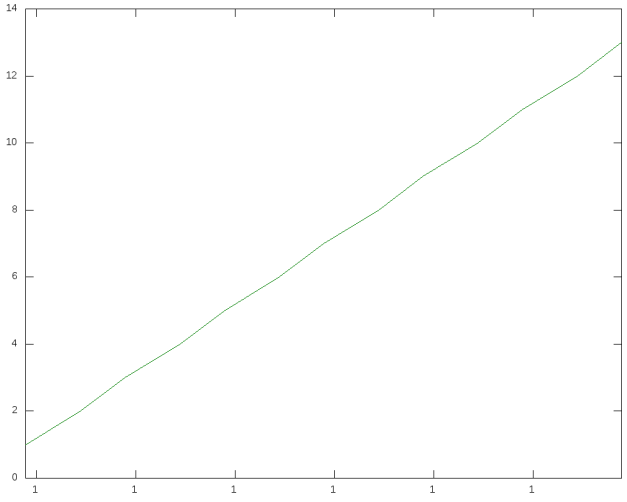
[4] Code used: ex = j; x = 1+0*10^ex:10^ex:1+12*10^ex; y = 1:13;[b,s,r] = ols(y,x);plot(x,y,x,x*b);axis([1 x(size(x)(2)) 0 14]);

Figure 4: $(1 + i/10^{14}, i + 1)$ in Octave



Only when $j = 15$ does Octave begin to reveal the underlying discrete nature of floating point.

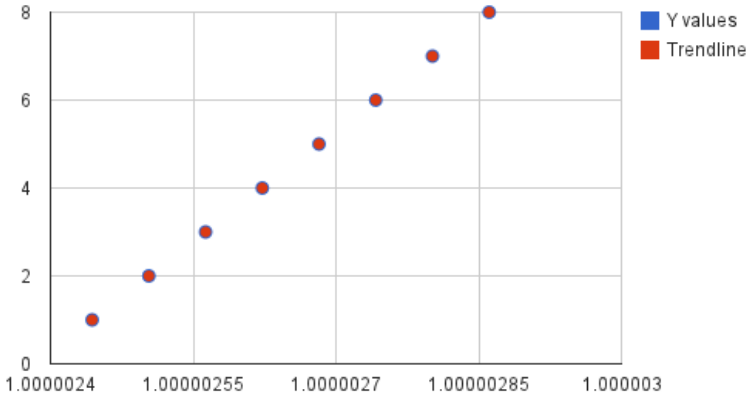
Figure 5: $(1 + i/10^{15}, i + 1)$ in Octave



At $j = 16$, Octave finally has an error and displays nothing, but there is no crash, and the failure to produce an answer makes sense, as 10^{-16}, the spacing of the x coordinates, is $\approx \epsilon/2$, where ϵ is provided in Octave as eps.

For better comparison with Kaplan's work on Excel, we also examined data series' of the form $(1 + (i + 40)/2^{-j}, i)$, where $1 \leq i \leq 8$. Kaplan found that Excel's Trendline function only performed well with $j \leq 25$, with bad error up to $j \leq 27$, and various degrees of failure past that.
Google Spreadsheet preforms relatively well ($1 - R^2 \approx 10^{-10}$ and 5 correct significant digits) up to $j = 19$. Up to $j = 24$, the error is still not noticable in a chart, and $1 - R^2 < 10^{-10}$, but even that accuracy is gone at $j = 25$.

Figure 6: $(1 + (40 + i)/2^{24}, i)$ in Google Spreadsheet



When $j = 25$, the error has become solidly visible. $R^2 = 0.997914698999419$, and there is not even always 1 correct significant digit in the trendline.

A little further, when $j = 27$, $R^2 is down to 0.928172702124667$, and there is no mistaking the divergent lines in the chart.

At $j \geq 28$, we get our now familiar "Oops", and the page reloads.

Figure 7: $(1 + (40 + i)/2^{25}, i)$ in Google Spreadsheet

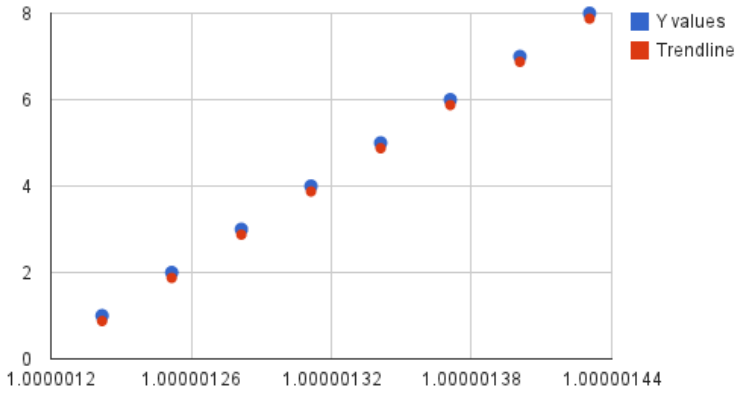
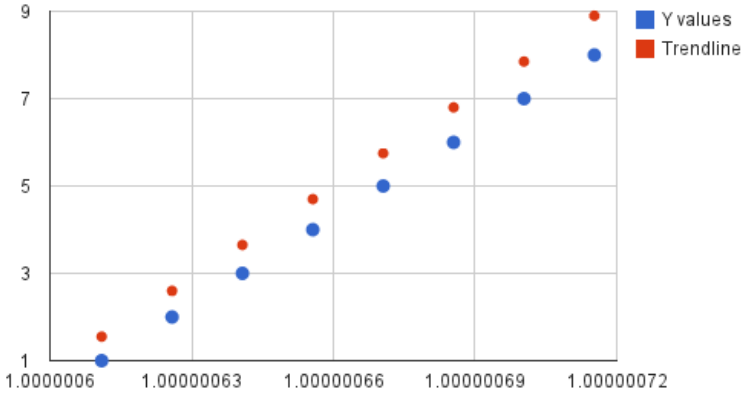


Figure 8: $(1 + (40 + i)/2^{27}, i)$ in Google Spreadsheet



Conclusion

In this paper, we explored Google Spreadsheet's floating point numbers, and investigated the built-in TREND function. Google Spreadsheet's TREND is significantly less accurate than the Excel function it is modeled after. Where Excel had good performance up to a slope of $10^2 5^5$, Google Spreadsheet had significant error by that time. When compared to functions from mathematical languages like Octave or Matlab, Google's showing is even poorer. However,

[5] Kaplan 2010 http://cs.nyu.edu/courses/spring12/CSCI-UA.0421-001/kaplanproject.pdf

Google Spreadsheet has none of the sleight of hand that Excel uses to conceal the nature of floating point, and no magical parentheses [6]. are needed to get functions to evaluate. On the other hand, when fatal errors do occur, Excel and Octave, and presumably most other programs, are able to display error messages, while Google seems to have decided that pretending the input which caused the error never occured is the most prudent route. Google should make their Spreadsheet more robust to ill-conditioned input, and display informative warnings and errors, rather than an "Oops", and a crash. Ideally, they would improve the accuracy of their algorithms, but it is more important to let users know when a dangerous numerical situation has arisen, perhaps with Octave like warnings of "a division by 0", or "rank deficient data". For now, users who need robustness or whose data may be ill-conditioned should steer clear of Google Spreadsheet's linear regression functionality.

[6] How Futile are Mindless..., Errors designed not to be found: http://www.cs.berkeley.edu/ wkahan/Mindless.pdf