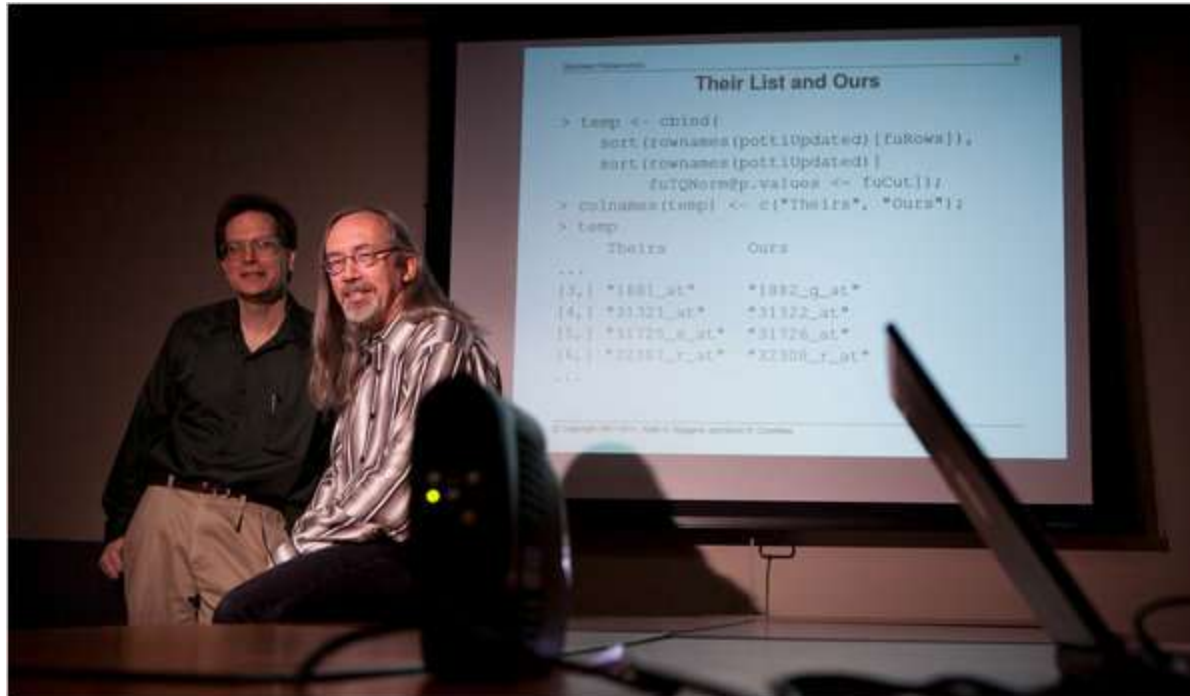


Importance of Open Access to Research Data: A Case in Point

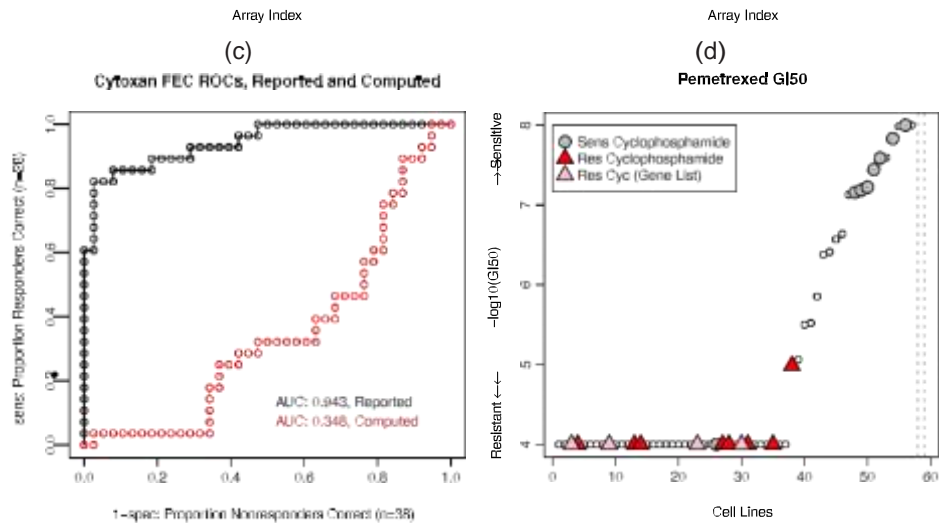
Helene Z Hill
NJ Medical School
Newark, NJ

How Bright Promise in Cancer Testing Fell Apart



Keith Baggerly, left, and Kevin Coombes, statisticians at M. D. Anderson Cancer Center, found flaws in research on tumors.

NY Times 7/7/11



(c) ROC curves for the single best set of drug predictions: cyclophosphamide for FEC. The reported curve has an AUC of 0.943, indicating extremely good prediction. Our best approximation has an AUC of 0.348, indicating if anything performance worse than chance.

[Retraction Watch](#)

Tracking retractions as a window into the scientific process

Anesthesiologist Fabricates 172 Papers

A researcher in Japan faked patient data on nearly 200 studies over the past 2 decades, according to an investigating committee.

J Med Ethics doi:10.1136/jme.2010.040923

•Research ethics

Retractions in the scientific literature: is the incidence of research fraud increasing?

1. [R Grant Steen](#)

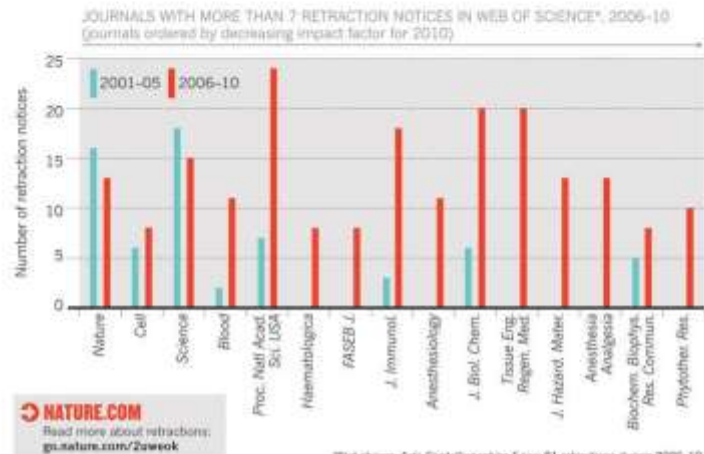
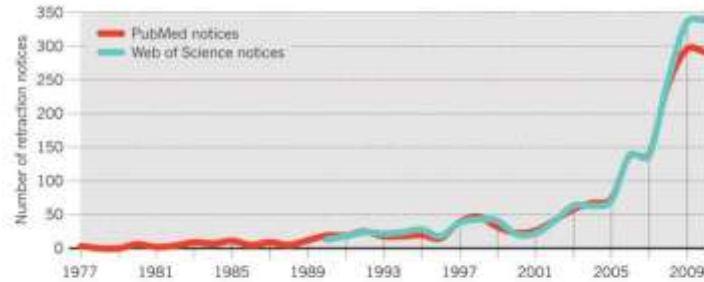
**Parkinson's Researcher Fabricated Data
Neuroscientist Mona Thiruchelvam agrees to retract two studies linking neurodegeneration to pesticides.**

By Hayley Dunning | June 29, 2012

[University of Michigan psychologist resigns following concerns by statistical sleuth Simonsohn: Nature](#)

RISE OF THE RETRACTIONS

In the past decade, the number of retraction notices has shot up 10-fold (**top**), even as the literature has expanded by only 44%. It is likely that only about half of all retractions are for researcher misconduct (**middle**). Higher-impact journals have logged more retraction notices over the past decade, but much of the increase during 2006-10 came from lower-impact journals (**bottom**).

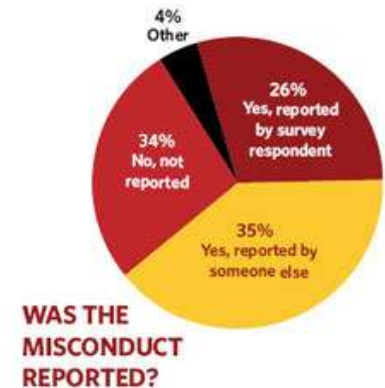
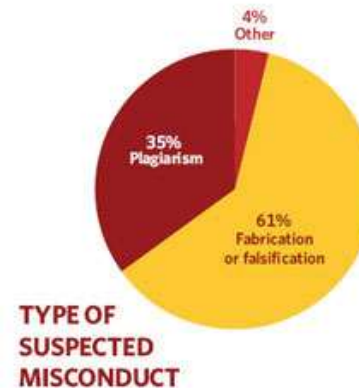
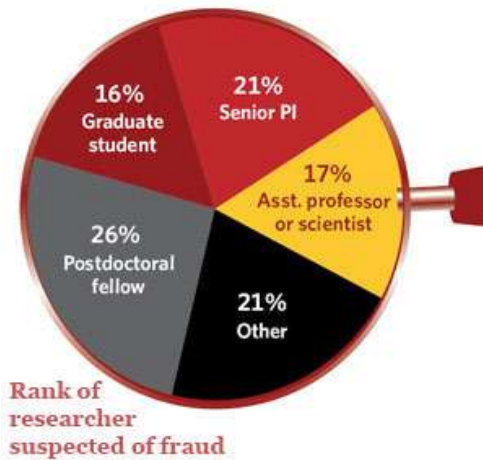
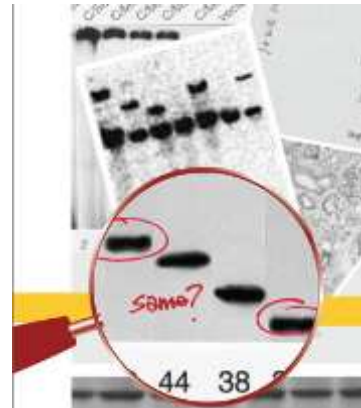


NATURE.COM
Read more about retractions:
go.nature.com/2awoek

*Not shown: Acta Crystallographica E saw 81 retractions during 2006-10.

A Gawrylewski Fixing Fraud *The Scientist* 23: 67 (2009)

Images are the easiest to spot



*According to a 2008 Gallup poll sent to 2,296 researchers receiving NIH grants

Fanelli D (2009) How Many Scientists Fabricate and Falsify Research? A Systematic Review and Meta-Analysis of Survey Data. **PLoS ONE 4(5):** e5738. doi:10.1371/journal.pone.0005738

“A pooled weighted average of 1.97% (N = 7, 95%CI: 0.86–4.45) of scientists admitted to have fabricated, falsified or modified data or results at least once –a serious form of misconduct by any standard– and up to 33.7% admitted other questionable research practices. In surveys asking about the behaviour of colleagues, admission rates were 14.12% (N = 12, 95% CI: 9.91–19.72) for falsification, and up to 72% for other questionable research practices. “

“...misconduct was reported more frequently by medical/pharmacological researchers than others.”

Source of the Data

- *Qui tam* is a suit filed by a private individual – relator – on behalf of the federal government charging violation of the False Claims Act
- An NCI grant application and its renewal (9 years of funding, \$2.5 million) were at stake
- Discovery is the period during which documents are exchanged: PDF files of scans of the laboratory notebooks from 1995 – 2003 were provided by the defense
- The Court ruled in favor of the defendants. The ruling was sustained on appeal
- The Courts decisions had nothing to do with the science, only with the timing of the alleged fraud

Importance of the Studies We Analyzed

in the PI's own words

“The outcome of this research is expected to have a major impact on understanding and predicting biological response of tumor and normal tissue to nonuniform distributions of radioactivity.”

The studies are “of importance to risk estimation in diagnostic nuclear medicine and radiation protection (radon, radiological terrorism), as well as clinical outcome in therapeutic nuclear medicine”

“In....diagnosis, the risk of radiation insult can ...be drastically underestimated and ... lead to increased risk of inducing cancer. In contrast, patients can be over or under treated in radionuclide therapy of cancer”

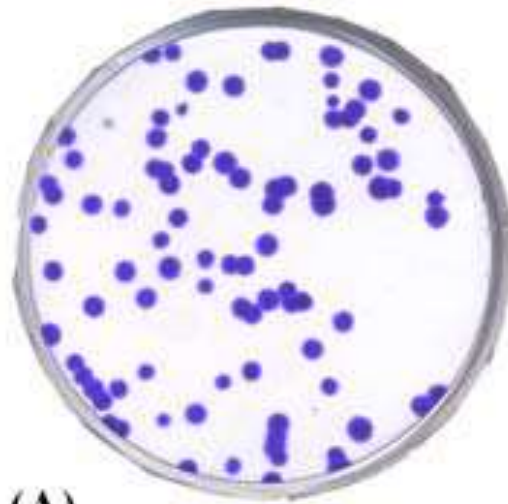
Data We Analyzed

- Numbers recorded by hand taken from the screen of a Coulter ZM Counter
 - In triplicate: 3 independent samples from an original cell suspension were counted
- Stained colonies on 60mm tissue culture dishes were counted by marking each on the underside of the dish
 - In triplicate: 3 independent samples from the same original suspension diluted appropriately were counted
- There were 5155 Coulter counts in 171 experiments and 3501 colony counts in 114 experiments performed by post-doctoral fellow AB
- There were 2759 Coulter counts in 99 experiments and 1556 colony counts in 59 experiments performed by 7 other individuals in the laboratory
- There were 674 Coulter ZM counts in 27 experiments provided by 2 outside laboratories

Coulter ZM



Chinese hamster lung V79 colonies



(A)

Terminal Digit Analysis

Accountability in Research, 9: 75-92, 2002
Copyright © 2002, Taylor & Francis
0898-9621/02 \$12.00 + .00
DOI: 10.1080/08989620290009530



Terminal Digits and the Examination of Questioned Data

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Accountability in Research, Vol. 4, pp. 31-55
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Gordon and Breach Science Publishers SA
Printed in Malaysia

Data Fabrication: Can People Generate Random Digits?

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Our Working Hypotheses

- Terminal digits of Coulter counts should be uniformly distributed (equal probability for each of the 10 digits 0, 1, 2, etc)
- Terminal digits of colony counts should likewise be uniformly distributed
- In Coulter counts, the second rightmost digit should be equal to the rightmost digit 10% of the time

Examples of Triplicate Coulter Counts

These were reported in an experiment by laboratory investigator AB and in an experiment by another laboratory investigator (OI). **The terminal digits are shown in bold.**

Sample #	AB Triplicate Counts			OI Triplicate Counts		
1	577	592	563	89	97	86
2	611	607	653	331	316	329
3	581	593	617	378	330	375
4	633	645	619	333	404	367
5	511	537	549	396	382	408
6	544	562	573	342	331	344
7	666	672	693	340	349	344
8	601	572	633	325	347	304
9	511	529	541	315	291	283
10	532	555	562	307	339	323
11	513	549	562	285	314	323
12	562	539	547	260	262	284
13	560	542	522	361	315	298
14	680	669	671	355	324	356

The Terminal Digit Counts from Above Table and The Chi-Squared Probability of Uniform Distribution

The chi-squared goodness of fit was determined in Microsoft Excel (9 degrees of freedom) for the digit frequencies of AB and OI compared with the control uniform distribution.

Digit	0	1	2	3	4	5	6	7	8	9	Total	Chi Sq	Chi sq p for uniform
AB Freq	2	7	9	8	1	2	2	5	0	6	42	21.8	0.0095
OI Freq	3	4	3	4	7	6	4	4	3	4	42	3.7	0.93
Ctrl Freq	4.2	4.2	4.2	4.2	4.2	4.2	4.2	4.2	4.2	4.2	42		

Examples of Triplicate Coulter Counts

These were reported in an experiment by laboratory investigator AB and in an experiment by another laboratory investigator (OI). The terminal digits are shown in **bold**. **The terminal duplicates are shown in red**. There are 10 doubles in AB's samples (23.8%) and 4 in the other investigator's samples (9.5%).

Sample #	AB Triplicate Counts			OI Triplicate Counts		
1	577	592	563	89	97	86
2	611	607	653	331	316	329
3	581	593	617	378	330	375
4	633	645	619	333	404	367
5	511	537	549	396	382	408
6	544	562	573	342	331	344
7	666	672	693	340	349	344
8	601	572	633	325	347	304
9	511	529	541	315	291	283
10	532	555	562	307	339	323
11	513	549	562	285	314	323
12	562	539	547	260	262	284
13	560	542	522	361	315	298
14	680	669	671	355	324	356

The Terminal Digit Counts from Above Table and The Chi-Squared Probability of Uniform Distribution

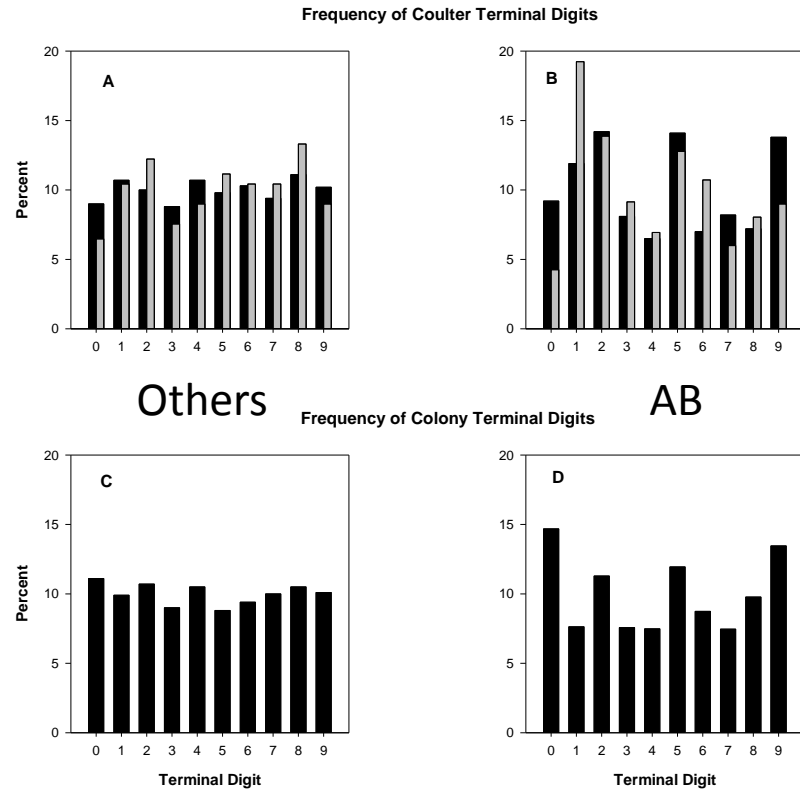
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Digit	0	1	2	3	4	5	6	7	8	9	Total	Chi Sq	Chi sq p for uniform
AB Freq	2	7	9	8	1	2	2	5	0	6	42	21.8	0.0095
OI Freq	3	4	3	4	7	6	4	4	3	4	42	3.7	0.93
Ctrl Freq	4.2	4.2	4.2	4.2	4.2	4.2	4.2	4.2	4.2	4.2	42		

Terminal Digit Analysis

In Figure A, black bars, the distribution of terminal digits of 2759 Coulter counts recorded by 7 members of the laboratory are shown. The grey bars show the distribution of terminal doubles in this set. Figure B shows similar results for 5155 Coulter counts recorded by post-doc AB.

Figure C shows the distribution of the terminal digits of 1556 colony counts recorded by 7 members of the lab. Figure D shows the 3501 colony terminal digit distribution for post-doc AB. Note the similarity of the patterns in B and D.



Terminal Digit Analysis of Coulter and Colony Counts

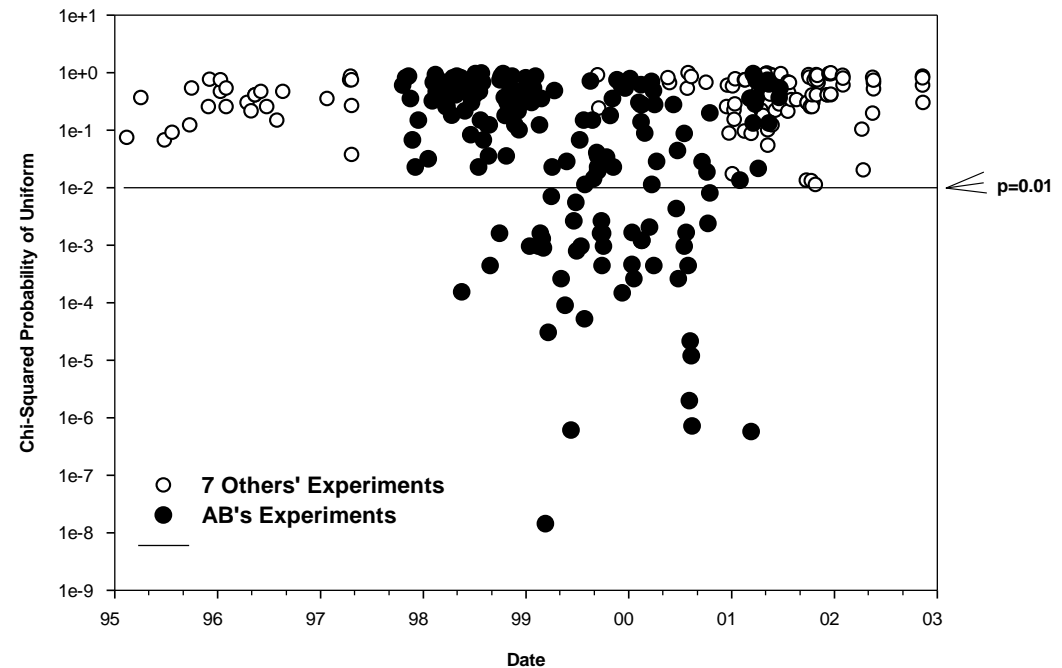
Chi Squared p-values were calculated using the program available in Excel by comparing the actual digit counts to the uniform distribution. The Coulter and colony distributions by others in the lab and from outside are quite likely due to chance. AB's distributions are extremely unlikely to be due to chance.

Type	Investigator	Digit										Total	Chi-sq	P-value
		0	1	2	3	4	5	6	7	8	9			
Coulter	AB: 171 experiments	472	612	730	416	335	725	362	422	370	711	5155	456.4	1.22x10⁻⁹²
Coulter	AB: terminal doubles	27	124	88	58	43	81	68	38	52	57	636	113.1	3.40x10⁻²⁰
Coulter	7 Others: 99 experiments	249	294	276	244	296	270	284	258	306	282	2759	13.9	0.13
Coulter	7 Others: terminal doubles	18	29	34	21	25	31	29	29	37	25	278	10.6	0.30
Coulter	Outside lab 1: 11 experiments	28	34	29	24	27	36	44	33	26	33	314	9.9	0.36
Coulter	Outside lab 2: 17 experiments	34	38	45	35	32	42	31	35	35	33	360	4.9	0.84
Colonies	AB 114 experiments	514	267	395	265	262	418	306	261	342	471	3501	228.4	3.56x10⁻⁴⁴
Colonies	7 Others: 59 experiments	173	154	166	140	163	137	147	156	163	157	1556	7.6	0.57

Coulter Chi-Squared p-Values Over Time

AB joined the lab in October, 1997 and left in July, 2001. For 45 experiments the hypothesis of uniformity would be rejected at the 0.01 level, a stringent testing condition. All of the improbable results were seen in experiments performed by AB.

Coulter Chi-Squared Probability of Uniformity Over Time



Colony Counts ~ PDF Copy from AB's Notebook

- We noticed that the **rounded** average of the 3 counts appeared unusually often as one of the triplicate counts in AB's colony records. In this example, it appears in 9 of the 10 samples (high-lighted in aquamarine). SF stands for Surviving Fraction, assumed to be 1.00 in Sample 1.2

Expt #: 2

Date: 02/22/99

Colony Counts and Survival Fraction

Tube, dilution	Colony 1	Colony 2	Colony 3	Avg Colony 62.2	SF
1.2	130	149	142	140.33	-
2.2	131	137	143	137.0	0.9762
3.2	123	131	138	130.66	0.9311
4.2	128	134	140	134	0.9548
5.2	125	130	136	130.33	0.9287
6.3	115	126	137	126	0.089
7.2	17	20	29	20.33	0.1484
8.2	29	35	41	35	0.2678
9.2	62	70	54	62	0.4626
10.2	70	79	62	70.33	0.5396

444
0
20
40
80
100

The Mid-Ratio

Non Sequitur by Wiley



The Mid-Ratio

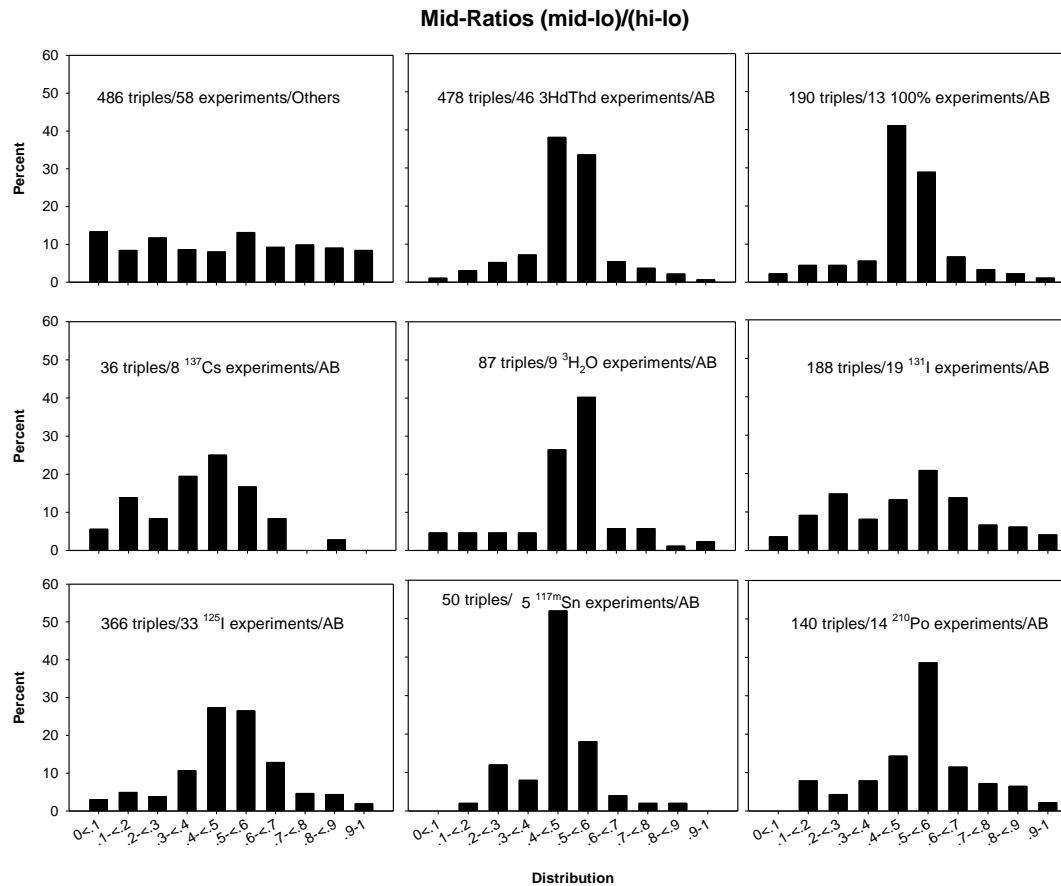
- The Mid-Ratio is the difference between the middle colony count and the lowest colony count divided by the highest count minus the lowest (for colony counts in rank order, $(b-a)/(c-a)$). When one of the counts in the triple is close to the triple average, its mid-ratio will be close to 0.5
- E.g. for the triple 17, 27, 35: $MR = (27-17)/(35-17) = 0.55$

The Mid-Ratio and the Appearance of the Rounded Average in Some of the Triple Colony Counts

Two experiments, one performed by AB, the other by another investigator (OI) in the lab. The rounded average, shown in aquamarine, appears in 9 of AB's 10 samples, but does not appear at all in the OI's samples. AB's mid-ratios are all close to 0.5. OI's are much more spread out.

Sample #	AB Triplicate Counts			Average	Mid-ratio (b-a)/(c-a)	OI Triplicate Counts			Average	Mid-ratio (b-a)/(c-a)
1	130	149	142	140.33	0.63	92	111	119	107.33	0.70
2	131	137	143	137	0.5	78	85	74	79	0.36
3	123	131	138	130.66	0.53	142	126	120	129.33	0.27
4	128	134	140	134	0.5	120	129	121	123.33	0.11
5	125	130	136	130.33	0.45	64	68	79	70.33	0.27
6	115	126	137	126	0.5	92	101	78	90.33	0.61
7	17	20	24	20.33	0.43	74	62	94	76.67	0.38
8	29	35	41	35	0.5	89	69	67	75	0.091
9	62	70	54	62	0.5	85	87	97	89.67	0.17
10	70	79	62	70.33	0.47	71	58	55	61.33	0.19

Mid Ratios in AB's Experiments with Different Isotopes



Statistical Analysis Of The Number Of Triples That Contain Their Rounded Mean: Calculating the Probability That One of the Triples Will be Equal to the Rounded Mean of that Triple

- The gap between the highest and lowest of the colony counts is variable, so the probability that one of the triples will equal the mean varies with the size of the gap. For example
 - The triple 8,x,10 has a gap of only 2 and the probability that the middle number will equal the rounded mean is 0.33
 - for x=8, the mean is 8.7 (rounds to 9)
 - for x=9, the mean is 9
 - for x=10, the mean is 9.3 (rounds to 9))
 - The triple 21,x,49 has a gap of 18. The middle is 35 and the probability that x=35 is 1/36 (about 0.28)
- Worse than that, the probability also depends on whether the gap is odd or even

The Probability of Hitting the Rounded Mean Depends on Whether the Gap is Odd or Even

- For triple colony counts $10, x, 20$, the gap is 10 (even) and there is 1 middle value, 15
 - For $x=14$ (triple is $10, 14, 20$), the mean is 14.7 (rounds up to 15)
 - For $x=\underline{15}$ (triple is $10, \underline{15}, 20$), the mean is 15
 - For $x=16$ (triple is $10, 16, 20$), the mean is 15.3 (rounds down to 15)

Only for $x=15$ is the rounded mean contained in the triple and the probability that the rounded mean is in the triple is $1/11$

- For triple colony counts, $10, x, 21$, the gap is 11 (odd) and there are 2 middle values, 15 and 16
 - For $x=\underline{15}$ (triple is $10, \underline{15}, 21$), the mean is 15.3 (rounds down to 15)
 - For $x=\underline{16}$ (triple is $10, \underline{16}, 21$), the mean is 15.7 (rounds up to 16) and the probability that the rounded mean is in the triple $2/12$

Statistical Analysis Of The Number Of Triples That Contain Their Rounded Mean

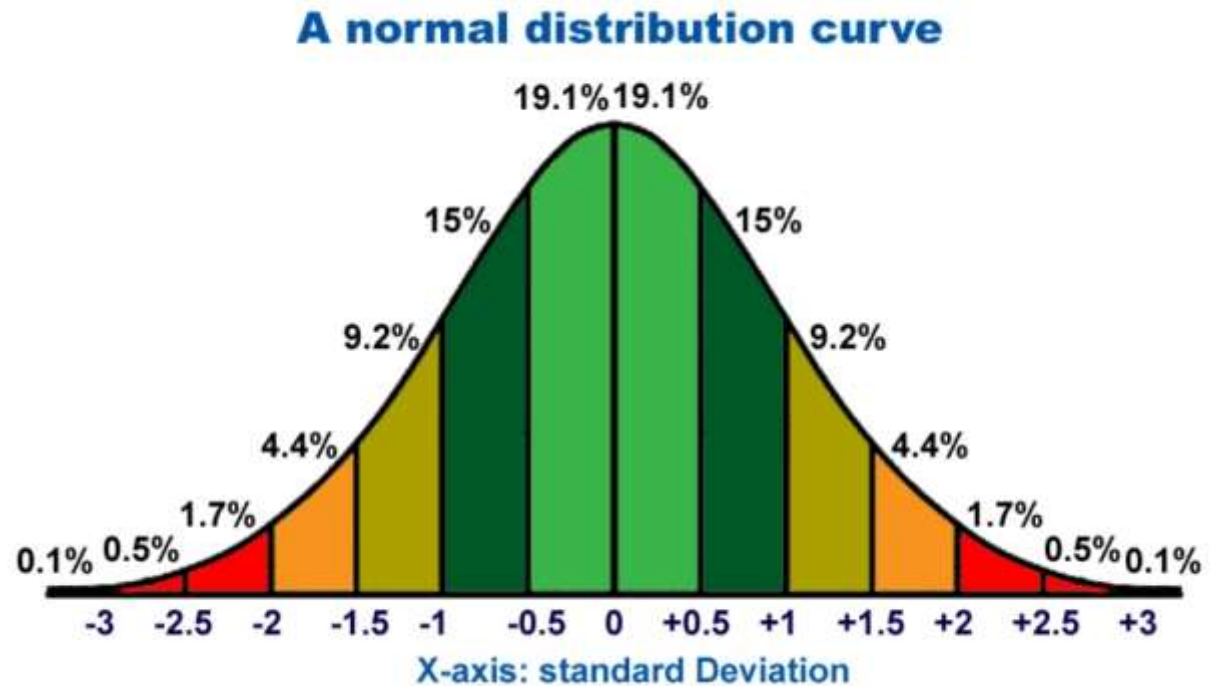
We can now calculate the probability of hitting the mean by chance for any given gap

- For even gap: probability of hitting the mean is $1/(1 + c - a)$ or $1/(1+gap)$
- For odd gap: probability of hitting the mean is $2/(1+c-a)$ or $2/(1+gap)$

Demonstration of Z-Scores

Z-Score

“a standard score [that] indicates how many standard deviations an observation or datum is above or below the mean.” *Wikipedia*



Calculating the Z-score and its probability

The values in column 4 were calculated for even gaps using the expression $1/(1+gap)$ and for odd gaps $2/(1+gap)$. The variances in column 5 were calculated from the values in column 4: variance = (col 4 value)*(1-col 4 value).

The Z-score is the actual number minus the expected number divided by the expected standard deviation.

	Sample #	Gap (c-a)	Column 4 Expected Number Rounded Means in Triples	Variances of Expected	Standard Deviation	Actual Number of Rounded Means in Triples	Z-Score for Actual Number	Probability of Z-Score
AB	1	19	0.100	0.090				
	2	12	0.077	0.071				
	3	15	0.125	0.109				
	4	12	0.077	0.071				
	5	11	0.167	0.139				
	6	22	0.043	0.042				
	7	7	0.250	0.188				
	8	12	0.077	0.071				
	9	16	0.059	0.055				
	10	17	0.111	0.099				
	Sum		1.086	0.934	0.967	9	8.19	<1x10 ⁻⁹
OI	1	27	0.071	0.066				
	2	11	0.167	0.139				
	3	22	0.043	0.042				
	4	9	0.200	0.160				
	5	15	0.125	0.109				
	6	23	0.083	0.076				
	7	20	0.048	0.045				
	8	22	0.043	0.042				
	9	12	0.077	0.071				
	10	16	0.059	0.055				
	Sum		0.917	0.806	0.898	0	-1.02	0.154

Z-Scores and Probabilities That Triples Contain Their Rounded Means Grouped by Isotope Experiments

Investigator	Isotope	# Experiments	# Triples	Expect # Triples Containing Rounded Means	Actual # Triples Containing Rounded Means	Z-Score for Actual Number	Probability of Z-Score or Higher
AB	^{117m} Sn	5	50	6.1	28	9.7	$\ll 1 \times 10^{-9}$
AB	²¹⁰ Po	14	140	19.7	67	12.0	$\ll 1 \times 10^{-9}$
AB	³ H ₂ O	9	90	13.2	28	4.8	1×10^{-6}
AB	¹³⁷ Cs External Beam	4	36	4.8	14	4.8	1×10^{-6}
AB	¹²⁵ I	33	375	67.9	174	18.4	$\lll 1 \times 10^{-9}$
AB	¹³¹ I	20	198	33.6	61	5.6	1×10^{-8}
AB	³ HdThd	44	478	65.0	301	35.1	$\llll 1 \times 10^{-9}$
OI	Various	59	534	86.4	95	1.2	0.123

Many of the survival experiments we examined involved compounds that contained different isotopes. In AB's hands, the actual number of triples that contain their rounded means far exceeds expectations based on expectations of randomness and independence. For other investigators (OI) using the same methods as AB, the frequency of rounded means is within the predicted expectation

RadioBiology

Non Sequitur by Wiley



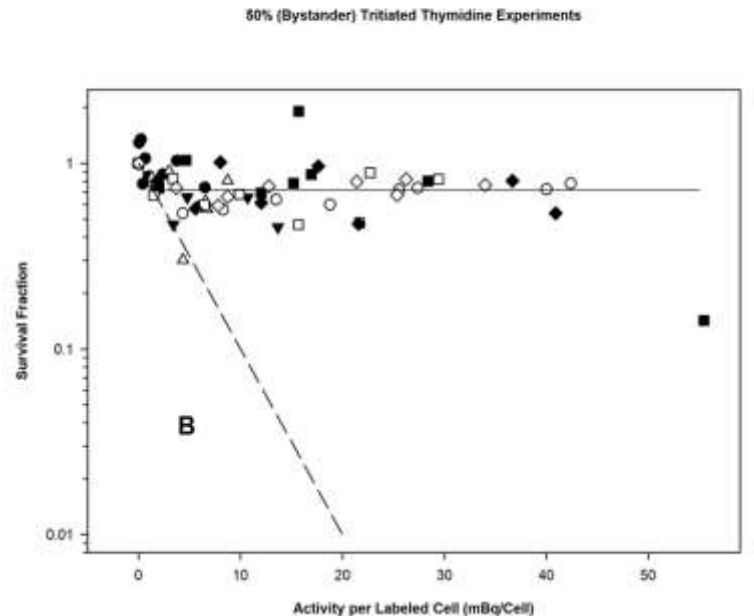
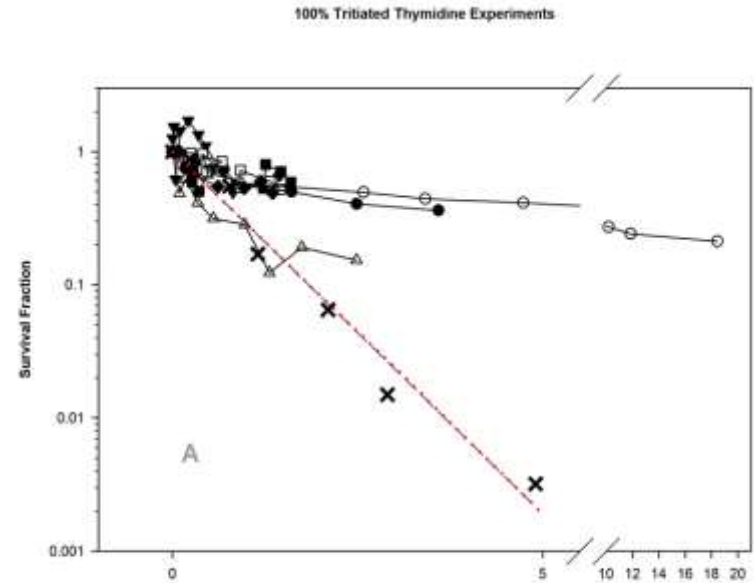
AB's $^3\text{HdThd}$ Survival Results Could Not be Replicated

The exponential survival s of V79 cells incubated with $^3\text{HdThd}$ reported in 2 publications could not be reproduced in 22 attempts to do so. In A, 100% of the cells were incubated with $^3\text{HdThd}$. The dashed line and X's represents results published in the two papers based on experiments performed by AB. The red line is the theoretical survival based on radiobiological principles. Ten attempts (symbols) to repeat AB's results were made by the PI and a second post-doctoral fellow.

Graph B shows 12 experimental attempts to repeat the bystander (50% experiments) results of AB (dashed line from the two papers). In these experiments, radioactive cells were incubated with non-radioactive cells.

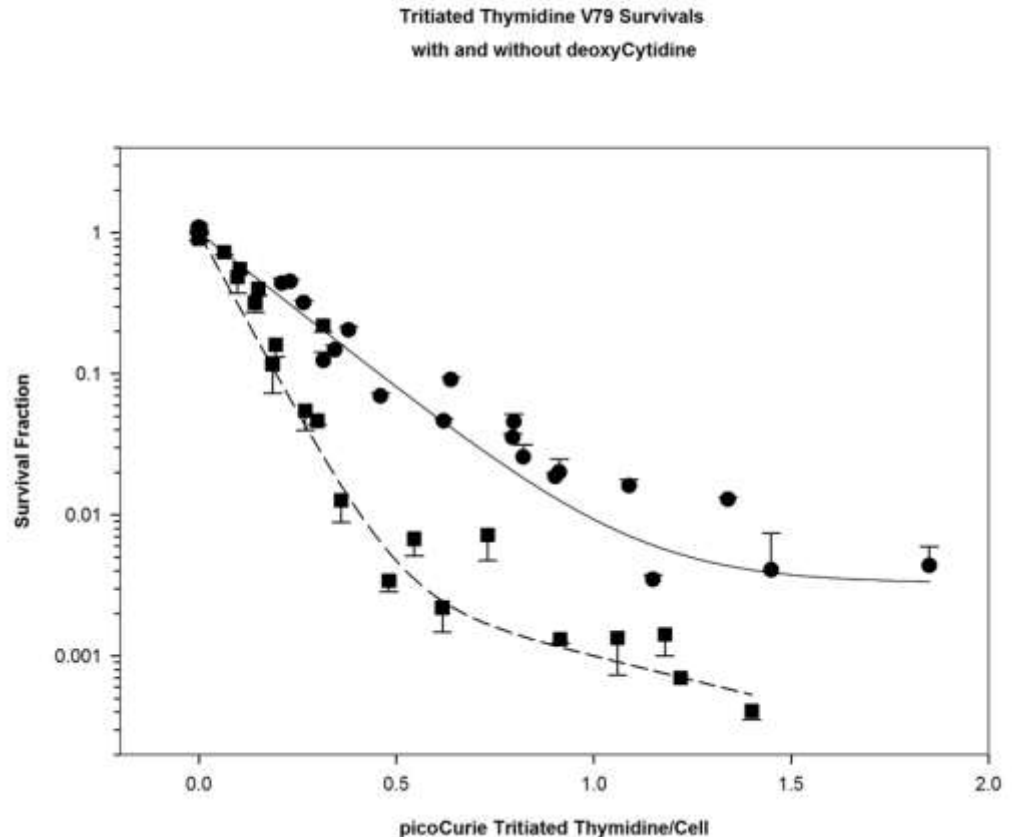
AB's results would argue for a bystander effect, the 12 experiments by the PI and a second post-doc (symbols) would argue against such an effect, at least under these conditions.

The results of these attempts to replicate are consistent with radiobiological literature that demonstrates $^3\text{HdThd}$ blocks the cell cycle unless deoxycytidine is present, which it was not in any of these experiments.

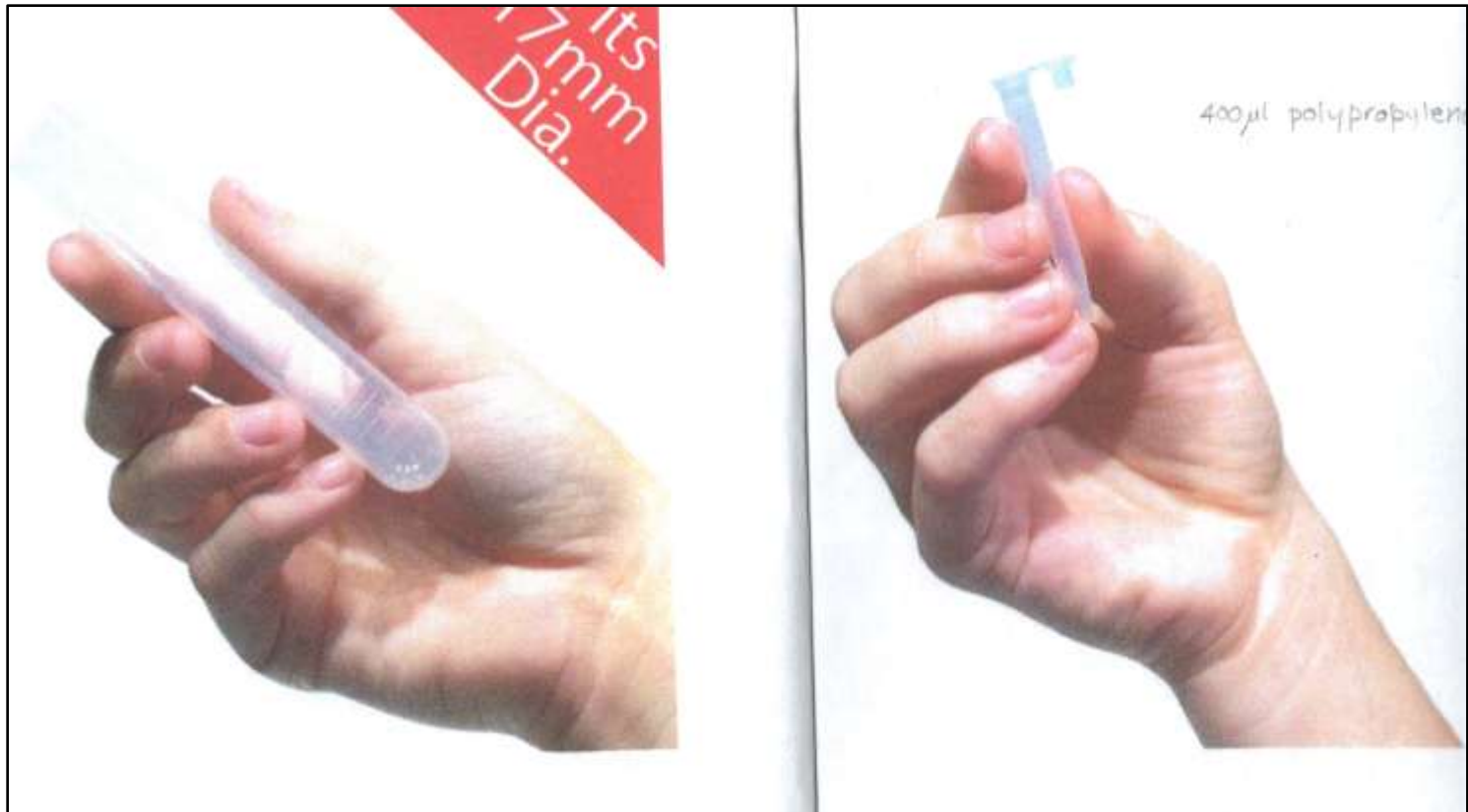


Deoxycytidine Reverses the Cell Cycle Blocking Effect of $^3\text{HdThd}$: No deoxycytidine was present in the experiments described in the preceding posters

These are 6 experiments performed by another investigator in the lab – 3 with deoxycytidine (squares) and 3 without (circles) that demonstrate the reversal by deoxycytidine of the cell cycle blocking effect of $^3\text{HdThd}$. No deoxycytidine was present in any of the other experiments described in these posters so AB's survival results are expected to be biphasic – but they are not.

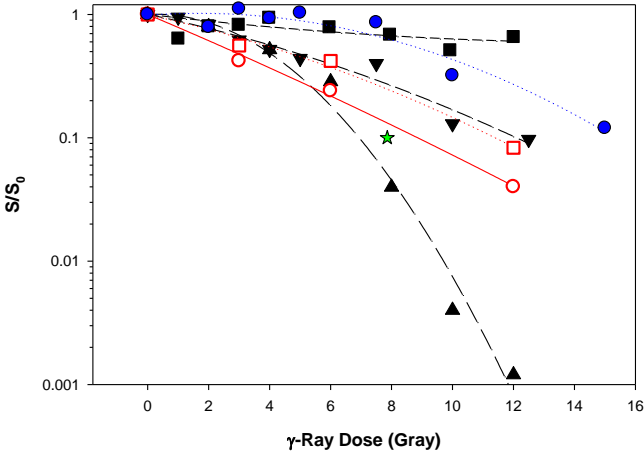


Culture Tubes



The Helena Tubes Are Hypoxic

Figure 7 - γ -Ray Survival of V79 Irradiated as Clusters or in Suspension



- ▲ B007921 Acute - Falcon Tubes
- B007921 Regression
- ▼ B007927/B007910-B007911 Clusters
- - - B007927 Regression
- B019656 Clusters (Marek)
- - - B019656 Regression
- B007894 Suspended
- Suspended Regression
- B007894 Clusters
- ⋯ Clusters Regression
- B002754-B002760 Clusters (Bogdan)
- ⋯ Bogdan Clusters Regression
- ★ V79 average Dose for 0.1 survival from literature

Summary

- The analysis reported here was only made possible because all the notebooks in the laboratory in question were made available through *subpoena*. Ordinarily, the raw data that underlies experiments are opaque to all but a few. We believe this was a rare and unique opportunity.
- We were fortunate to be able to compare data produced by several other individuals using the same instrument and techniques to that of a single individual, whose results alone could not be explained based on assumptions of randomness, uniformity or chance.
- We used the following tests, all of which are simple and could be put to use in any laboratory
 - Terminal digits of Coulter counts: AB's digit distributions diverged markedly from uniform, while others distributions were consistent with the assumption of uniformity*
 - Terminal digits of colony counts: AB's distributions were highly unlikely to be uniform or random, whereas others' colony terminal digit distributions are likely to have come about by chance*
 - Double terminal digits in the Coulter counts: AB's doubles are not likely to be due to chance, others' doubles are close to the expected 10%
 - Presence of the average as one of the triples in colony counts: AB's results had an inordinately high frequency of the rounded average occurring as one of the triplicate counts, results of others were consistent with the predicted frequency based on gap size
- The statistical analyses are bolstered by the inability to replicate AB's radiobiological results. The differentials in survivals are astounding – about 1000 fold in the 100% experiments, about 100 fold in the 50% (bystander) experiments

* Terminal digit analysis is recommended as a forensic tool on the USPHS Office of Research Integrity WebSite.

Conclusions

- The results of the statistical analyses reported here argue strongly for making raw data used in the production of scientific papers, research reports and grant applications available to all researchers, reviewers and granting agencies
- Our analyses also argue for sharing such raw data with other researchers in the field in order for them to understand their own, possibly unexpected, experimental results and/or allowing them to avoid performing unnecessary experiments
- The experiments involved in these studies were designed to alleviate a vexing problem in Nuclear Medicine – that of the non-uniform distribution of radioisotopes in the human body. These results were planned to be used in setting standards for allowable exposures for healthcare workers and to determine isotope doses to be used for diagnostic and therapeutic procedures
- Miscalculations based on these results could have serious consequences for patients and workers in nuclear medicine: over-estimating doses could lead to tissue and organ damage and even to cancer; under-estimating doses could lead to misdiagnoses and lack of therapeutic efficacy

Thanks

- Mike Falvo, PhD for listening to my story
- George J Hill, MD, DLitt for his constant support and encouragement
- Joel Pitt, PhD My expert Statistician
- Mike Robbins, PhD My expert Radiobiologist
- Drummond Rennie, MD
- Harold Sox, MD
- Miriam Wahl, PhD

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Have Comments?
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Truth even unto its innermost parts

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This Website Exposes a Scientific and Medical Cover Up

Welcome to my WebPage. Please understand that it is still under construction and many of the links are yet to be completed.

In October of 2004, I filed a case for *qui tam* in the Federal District Court of Newark charging my university and two scientists with whom I had been working with violation of the False Claims Act. I lost the case. There are a number of posts on the web regarding that case, all of which present one side of the story: that of the defendants. This webpage is my answer to those posts, and it shares a story themed around *medical cover up*. On it, I tell all who visit it something about me, my family, my scientific career, but most of all, I will present the information that I acquired during the case in order for others to decide whether I was right or wrong.

Most important, it will become apparent that the whistleblower has no right to appeal unfavorable decisions at the university level – although the defendant can appeal. Secondly, I would have had no access to the original data that I eventually analyzed had it not been for the filing of the *qui tam* case, so I argue strongly that raw data on which papers, reports and grant applications are based must be freely available for scrutiny by both readers and referees; a practice that would greatly promote **scientific accountability**.



Helene Z. Hill, PhD



The data here analyzed were involved in the following publications:

Bishayee, A., H. Z. Hill, et al. (2001). "Free radical-initiated and gap junction-mediated bystander effect due to nonuniform distribution of incorporated radioactivity in a three-dimensional tissue culture model." *Radiat Res* **155**(2): 335-344.

Bishayee, A., D. V. Rao, et al. (2000). "Protection by DMSO against cell death caused by intracellularly localized iodine-125, iodine-131 and polonium-210." *Radiat Res* **153**(4): 416-427.

Bishayee, A., D. V. Rao, et al. (1999). "Evidence for pronounced bystander effects caused by nonuniform distributions of radioactivity using a novel three-dimensional tissue culture model." *Radiat Res* **152**(1): 88-97.

Bishayee, A., D. V. Rao, et al. (2000). "Radiation protection by cysteamine against the lethal effects of intracellularly localized Auger electron, alpha- and beta-particle emitting radionuclides." *Acta Oncol* **39**(6): 713-720.

Bishayee, A., D. V. Rao, et al. (2000). "Marrow-sparing effects of ^{117m}Sn(4+)diethylenetriaminepentaacetic acid for radionuclide therapy of bone cancer." *J Nucl Med* **41**(12): 2043-2050.

Goddu, S. M., A. Bishayee, et al. (2000). "Marrow toxicity of ³³P-versus ³²P-orthophosphate: implications for therapy of bone pain and bone metastases." *J Nucl Med* **41**(5): 941-951.

Howell, R. W. and A. Bishayee (2002). "Bystander effects caused by nonuniform distributions of DNA-incorporated (125)I." *Micron* **33**(2): 127-132.

Howell, R. W., S. M. Goddu, et al. (1998). "Radioprotection against lethal damage caused by chronic irradiation with radionuclides in vitro." *Radiat Res* **150**(4): 391-399.

Howell, R. W. (2000-2006). 1R01CA083838-01A1 Effects of Nonuniform Distributions of Radioactivity.

Howell, R. W. (2006-2011). "1R01CA083838-06A1 Effects of Nonuniform Distributions of Radioactivity ".