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Red Ed	Dested: Thu Oct 22, 2009 6:22 am Post subject:					
	denis_berthier wrote:					
Joined: 06 Jun 2005 Posts: 858	Though systematically below, it is consistent with sampling errors.					
	"Consistently below" makes me think that the SE ratings for Allan's puzzles are subtly different to those for unbiased puzzles. We should expect them to be, given past experience.					
	Interesting that Allan's result for 32s is slightly above what eleven and I reported.					
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denis_berthier	Dested: Thu Oct 22, 2009 6:33 am Post subject:					
	Red Ed wrote:					
Joined: 19 Jun 2007 Posts: 970	given past experience.					
Location: Paris, France	A reference which was only a summary of my anterior reports 😁					
	Red Ed wrote:					
	Interesting that Allan's result for 32s is slightly above what eleven and I reported.					
	Interesting but, considering sampling errors, no more meaningful than the above difference.					
	What I'd like to see now is the #clues distribution if Allan can run his generator with no preset number of clues.					
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Red Ed	D Posted: Thu Oct 22, 2009 7:18 am Post subject:
Joined: 06 Jun 2005 Posts: 858	Actually the point of referencing that post, which goes beyond what you had previously observed, was to suggest that a similar distributional analysis might be worthwhile with new data - <i>i.e.</i> Allan's puzzles just produced <i>versus</i> a set of puzzles produced in an unbiased fashion (both for the same fixed number of clues). It would tell us so much more than the mean & sd alone.
	But if it makes your day to be reminded of your past work, then I'm happy to have obliged 😅
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Red Ed	Dested: Thu Oct 22, 2009 7:33 am Post subject:
Joined: 06 Jun 2005 Posts: 858	btw, mean SER on 1400+ 31-clue puzzles produced by the supersets method is 6.94; no time to produce a measure of confidence for that now though.
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Allan Barker	Denosted: Thu Oct 22, 2009 4:52 pm Post subject:
	Denis Berthier wrote:
Joined: 21 Feb 2008 Posts: 362 Location: Bangkok	A quick SER computation gives: We can see a very strong upward trend.
	Although not enough for any statistical confidence, I now have 5 <u>37-clue</u> puzzle with average SER of 8.0.
	Denis Berthier wrote:
	In the range where comparisons are possible, it is closer to the real values than any upward or top-down generator. Though systematically below, it is consistent with sampling errors.
	That's interesting. Although the routine starts by removing 81-N candidates fror a grid, the grid is then randomized during the convergence process (clues change). However, there would still be bias from this remaining process.
	Red Ed wrote:
	"Consistently below" makes me think that the SE ratings for Allan's puzzles are subtly different to those for unbiased puzzles.
	Yes, I assume that is the bias I mention.
	Denis Berthier wrote:
	What happens if you don't a preset the number of clues?
	I don't know (!) If the number of clues can change in a search for fewer non-

	essential clues, it could revert to a top down generator. As Ed Red mentioned though, comparing distributions would be interesting. I'll see if there is a way to get a distribution while retaining the essence of the method.				
Back to top	Last edited by Allan Barker on Sat Oct 24, 2009 1:10 am; edited 1 time in total				
Red Ed	D Posted: Thu Oct 22, 2009 5:31 pm Post subject:				
Joined: 06 Jun 2005 Posts: 858	Allan, for distributions, it would suffice to just produce a list of (SER,#clues) pairs; then we can cut it any way we like for distribution info - e.g. fixed number of clues (e.g. is your method more prone to fish than unbiased generation?), or fixed method (e.g. how does fishiness depend on #clues in your method?).				
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eleven	D Posted: Fri Oct 23, 2009 8:07 am Post subject:				
Joined: 10 Feb 2008 Posts: 549	OT: I was interested, how the {-n+m} performs to find high clue puzzles. I took Red Eds big 32 cluster as starting point and used gsf's program. Code:				
	$30 \ 32's: \min 2, \max$ 8.3, average 4.88 $\{-1+2\}, \{-1+1\} \qquad -> 9/ \ 41 \ 33's: \min 2.5, \max$ 7.2, average 6.05 $\{-1+2\}, 4x\{-1+1\} \qquad -> 5/ \ 66 \ 34's: \min 2.6, \max 9,$ average 6.99 $\{-1+2\}, \{-2+2\}, 4x\{-1+1\} \qquad -> 6/156 \ 35's: \min 2.6, \max$ 9.1, average 7.91 $\{-1+2\} \qquad -> \ 33 \ 36's: \min 7.1, \max 9,$ average 7.91 $\{-1+2\} \qquad -> \ 1 \ 37 \qquad : 9.0$				
	This took about 11 hours. The {-1+2} for the 156 35's was extremely slow (more than 5 hours, where for 6 puzzles it needed 7 minutes), so there could be done a lot of optimization.				
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eleven	Dested: Fri Oct 23, 2009 8:33 am Post subject:				
Joined: 10 Feb 2008 Posts: 549	In a background process i have running the supersets method for 33 clues (from 24). The (intermediate) result for 0.5 mio random grids is				
	Code:				
	jct[0] = 508126 jct[1] = 37 jct[2] = 6 jct[3] = 2 jct[4] = 3 jct[7] = 1				

	jct[1000+] = 0
	508175 grids, 74 puzzles Estimated mean number of proper minimal 33s per grid = 8.704923e+08
	Estimated relative error = 17.93%
	74 puzzles, SER min 2, max 8.8, average 7.26
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Allan Barker	Deposted: Fri Oct 23, 2009 10:29 am Post subject:
	Red Ed wrote:
Joined: 21 Feb 2008 Posts: 362 Location: Bangkok	Allan, for distributions, it would suffice to just produce a list of (SER,#clues) pairs; then we can cut it any way we like for distribution info - e.g. fixed number of clues (e.g. is your method more prone to fish than unbiased generation?), or fixed method (e.g. how does fishiness depend on #clues in your method?).
	I don't think there is anything more fishy 😃 about this algorithm vs. any other (not including true unbiased generation). It treats the grid as a 9x9x9 cube and randomly chooses x,y,z pairs for swapping. The only decision is then whether the grid is valid, minimal, or non-minimal, which is necessary even for unbiased generation.
	Where the bias would creep in, and I'm sure it does, is the restriction to move one pair at a time. Now imagine choosing the number of pairs to swap each time, 1, 2, N, when N = the number of clues. When you change all N clues each time, you arrive at a completely unbiased generator (and its performance). If worthwhile, I could make a few data sets where the algorithm arrives at a minimal by changing different numbers of clues each step. You might then be able to measure the change in bias from one set to the next with your 3222 patterns test (?)
	Eleven , I am wondering how similar this is to your {+n,-m} method. Is there a good link that explains that method in detail?
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eleven	Dested: Fri Oct 23, 2009 11:47 am Post subject:
	Allan Barker wrote:
Joined: 10 Feb 2008 Posts: 549	Eleven , I am wondering how similar this is to your {+n,-m} method. Is there a good link that explains that method in detail?
	No link needed, {-n+m} means, remove n clues and add m. For a puzzle with N givens you have choose(N,n)*choose(81-N+n, m)*9^m possibilities to find minimals. I denoted the post as OT, because i had seen before, that this method produces rather strong bias.
	[Edit:]oops, i forgot the 9 ^m factor, the point is, that it is not restricted to a fixed grid.

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denis_berthier	D Posted: Sat Oct 24, 2009 5:54 am Post subject:
Joined: 19 Jun 2007	ANALYSIS OF SKEWNESS AND KURTOSIS
Location: Paris, France	Until now, I have given results for the mean, standard-deviation and distribution for various random variables of minimal puzzles: number of clues, SER, NRCZT.
	But there are two more variable of interest: skewness and kurtosis.
	 Skewness is a measure of the asymmetry of the distribution negative skew: the left tail is longer; the mass of the distribution is concentrated on the right of the figure; the distribution has relatively few low values. positive skew: the right tail is longer; the mass of the distribution is concentrated on the left of the figure; the distribution has relatively few high values. Details on Wikipedia (http://en.wikipedia.org/wiki/Skewness)
	Kurtosis (also called "excess kurtosis" to avoid ambiguities) is a measure of how much the distribution shape differs from that of a Normal distribution with the same mean and standard deviation as X. - lower kurtosis means more of the variance is due to frequent modestly sized deviations;
	 higher kurtosis means more of the variance is due to infrequent extreme deviations. Kurtosis can vary between - 2 and + infinity.
	Details on Wikipedia (http://en.wikipedia.org/wiki/Kurtosis)
	1) number-of-clues Kurtosis
	bottom-up generator (suexg1.4) : 0.026 top-down generator (suexg-x.x) : 0.007 (Allan's top-down = 0.014) controlled-bias generator (gsf suexg-cb) : 0.023 (suexg .suexg-cb : 0.013)
	Skweness bottom-up generator (suexg1.4) : 0.11 top-down generator (suexg-x.x) : 0.08 (Allan's top-down = 0.11) controlled-bias generator (gsf suexg-cb) : 0.08 (suexg .suexg-cb : 0.08)
	For all these generators, kurtosis and skewness of the number of clues are ~ 0 and the general shape of the number-of-clues distribution looks much like Normal.
	2) SER

	Kurtosis
	bottom-up generator (suexg1.4) : -1.08
	top-down generator (suexg-x.x) -1.36 : (Allan's top-down = -1.35)
	controlled-bias generator (gsf suexg-cb) : - 1.70 (suexg suexg-cb : -1.70)
	Skweness
	bottom-up generator (suexg1.4): 0.79
	top-down generator (suexg-x.x) : 0.60 (Allan's top-down = 0.60)
	controlled-bias generator (gsf suexg-cb) : 0.21 (suexg suexg-cb : 0.21)
	3) NRCZT
	Kurtosis
	bottom-up generator (suexg1.4) : 0.91
	top-down generator (suexg-x.x) : 0.27 (Allan's top-down = 0.34)
	controlled-bias generator (suexg .suexg-cb) : -0.46
	Skweness
	bottom-up generator (suexg1.4) : 1.24
	top-down generator (suexg-x.x) : 1.01 (Allan's top-down = 1.01)
	controlled-bias generator (suexg .suexg-cb :) 0.65
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denis_berthier	D Posted: Sat Oct 24, 2009 5:55 am Post subject:
Joined: 19 Jun 2007	KURTOSIS PROFILES
Posts: 970	
Location: Paris, France	For a random variable X other than the number of clues, kurtosis(X) can be
	computed for the subsamples of n-clue puzzles.
	We thus get a "kurtosis profile".
	Kurtosis-profile(SER)
	As shown by the tables below, all the generators we have studied have similar
	kurtosis profiles :
	 negative values in the range [22, 28]
	- a negative peak around 26/27 clues (notice: around the real mean number of
	clues and not around the mean number of clues for the generator)
	 positive values above 30 (when there are puzzles there)
	Moreover, the controlled-bias generator has exactly the same kurtosis profile
	whether the source of complete arids is suexa or asf :
	Code:
	Code: controlled-bias generator (gsf suexg-cb) :
	Code: controlled-bias generator (gsf suexg-cb) : 21 -0.65 (*)

23	-0.72
24	-1.18
25	-1.53
26	-1.73
27	-1.68
28	-1.31
29	-0.32
30	2.62 (*)
31	-0.64 (*)

Code:

control	lled-bias	generator	(suexg	Τ	suexg-cb)	:
21	-1.5 (*)					
22	-0.48					
23	-0.73					
24	-1.18					
25	-1.54					
26	-1.73					
27	-1.67					
28	-1.29					
29	-0.48					
30	1.23 (*)					

Code:

top-do	wn generator	(suexgx.x)	:
21	0.29 (*)		
22	-0.50		
23	-0.88		
24	-1.21		
25	-1.48		
26	-1.68		
27	-1.74		
28	-1.59		
29	-1.28		
30	3.64 (*)		

Code:

bottom-	up generator (suexg1.4) :
20	0.25
21	0.28
22	-0.28
23	-0.69
24	-1.08
25	-1.41
26	-1.66
27	-1.74
28	-1.54
29	1.57 (*)

Allan's montecarlo is interesting because it shows how the kurtosis evolves beyond 30 clues: there is a continuous increase, i.e. values away from the mean tend to be more and more frequent.

The skewness profile will show that these values away from the mean tend to be below the mean.

	Code:
	Allan's montecarlo :
	22 -0.36
	23 - 0.41
	24 -1.18 25 -1.55
	26 -1.71
	27 –1.72
	28 -1.44
	29 0.56
	30 0.20
	32 2.88
	33 5.38
	34 5.90
	35 8.48
	30 4.37 (^)
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denis_berthier	D Posted: Sat Oct 24, 2009 5:55 am Post subject: 🔍 quote 🐼 edit
Joined: 19 Jun 2007 Posts: 970	SKEWNESS PROFILES
Location: Paris, France	For a random variable X other than the number of clues, skewness(X) can be
	computed for the subsamples of n-clue puzzles.
	We thus get a "skewness profile".
	Skewness-profile(SER)
	Code:
	controlled-bias generator (gsf suexg-cb) :
	21 2.38 (*)
	22 0.001
	23 0.018
	25 0.14
	26 0.045
	27 -0.037
	29 -0.0056
	30 -0.0000 31 -1.3e-5 (*)
	There's almost no skewness

Code:	
top-dow	vn generator (suexgx.x) :
21	0.0003 (*)
22	0.037
23	0.15
24	0.24
25	0.15
26	0.03
27	-0.0006
28	-0.0009
29	-0.0001
30	-2e-5 (*)

Code:

bottom-	up generator (suexg1.4) :
20	0.0004
21	0.012
22	0.095
23	0.27
24	0.29
25	0.12
26	0.016
27	-6e-6
28	-0.0002
29	-4e-5 (*)

Although its skewness profile is different from all the others, Allan's montecarlo is interesting, because it can be computed upto 36 clues and it shows how puzzles with many clues (higher mean SER, higher kurtosis) tend to avoid having much higher complexity than the mean SER value for their number of clues: there is a decreasing skewness, i.e. the SER of puzzles with more clues tend to be below the mean SER of their number of clues.

Code:

Allan'	s montecarlo	:
22	1.12	
23	1.11	
24	0.74	
25	0.44	
26	0.20	
27	-0.11	
28	-0.51	

	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$		
	It is as if there was a barrier of complexity and, when the number of clues increases: - the mean complexity increases - the proportion of puzzles away from the mean increases but		
	 there are more and more puzzles far below this mean there are fewer and fewer puzzles far above this mean 		
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Red Ed	Dested: Sat Oct 24, 2009 7:53 am Post subject:		
	denis_berthier wrote:		
Joined: 06 Jun 2005 Posts: 858	For all these generators, kurtosis and skewness of the number of clues are \sim 0 and the general shape of the number-of-clues distribution looks much like Normal.		
	Nice. If I find the time this weekend then I'll redo my analysis of the fit of the Gamma distribution to this family. The analysis was done originally on "suexg .suexg-cb". Recall that after proposing Normal as a good approximation, I offered that Gamma might be a better fit: and for the parameters that I proposed, the skewness works out at 0.087 and the kurtosis at 0.011 good matches for your 0.08 and 0.013 respectively.		
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denis_berthier	D Posted: Sat Oct 24, 2009 8:40 am Post subject:		
	Red Ed wrote:		
Joined: 19 Jun 2007 Posts: 970 Location: Paris, France			
	If you can wait a few more days, I'll soon have a larger sample.		
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