

# Scientific Methods for Health Sciences: Applied Inference (HS851): Fall 2014

<http://www.socr.umich.edu/people/dinov/2014/Fall/HS851>

## Homework 1 Solutions

**Problem 1:** Suppose we have obtained the following measurements. Calculate the corresponding [power, specificity and sensitivity](#).

		Actual Condition	
		Absent (Ho is true)	Present (H1 is true)
Test Result	Negative(fail to reject Ho)	0.983	0.0025
	Positive (reject Ho)	0.0085	0.0055

Power:  $1 - (FN / (TP + FN)) = 0.6875$

Specificity:  $TN / (TN + FP) = 0.983 / (0.983 + 0.0085) = 0.9914271$

Sensitivity:  $TP / (TP + FN) = 0.0055 / (0.0055 + 0.0025) = 0.6875$

R calculations

```
dat <- matrix(c(0.983, 0.0025, 0.0085, 0.0055), byrow=T, nrow=2)
dat
```

```
colnames(dat) <- c('H0 true', 'H1 true')
rownames(dat) <- c('Neg test results', 'Pos test result')
power <- 1-dat[1,2]
specificity <- dat[1,1]/(dat[1,1]+dat[2,1])
sensitivity <- dat[2,2]/(dat[2,2]+dat[1,2])
```

**Problem 2:** Using [R](#) and [SOCR](#), generate 100 random observations from [uniform\(-1,1\)](#), [normal\(0,1\)](#) and [exponential\(1\)](#) distributions. Then Use the [SOCR QQ Data-Data Plot](#) to compare the paired samples, i.e., for each distribution, plot the quantiles of the R-generated sample against the SOCR generated sample. What do you expect and what do you see? Explain.

In all cases ([uniform\(-1,1\)](#), [normal\(0,1\)](#) and [exponential\(1\)](#) distributions) we expect to see the QQ plots to be Very linear with the scatter plot of the (R, SOCR) data evenly distributed over the range and close following the line bisecting the  $[0,1] \times [0,1]$  domain. There should not be S-shape (snake) patterns emerging, which would be indicative of differences in the quantile distributions (which is not expected as both software tools sample from the same distributions, albeit using different sampling methods). See Appendix for random samples.

**Using SOCR QQ Data to Data Plot** ([http://www.socr.umich.edu/html/cha/SOCR\\_Charts.html](http://www.socr.umich.edu/html/cha/SOCR_Charts.html))

Copy-paste pairs of data (randomly generated samples from R and SOCR) for each of the 3 distributions.

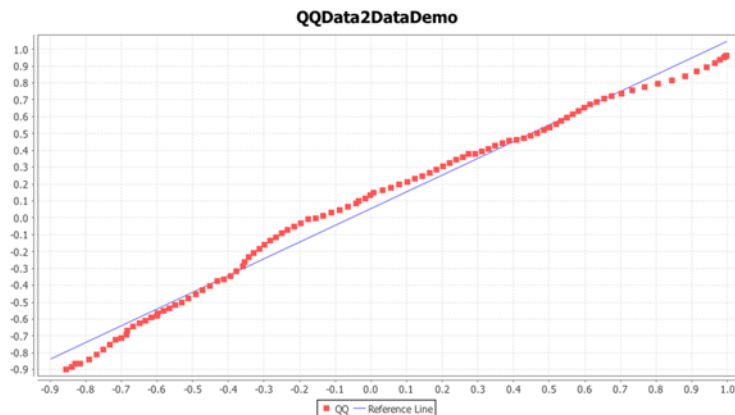
Summary:

QQ: SampleSize=101 Mean=85.0905 Median=100.67606 stdDev=29.91073 Skewness=-1.31116 Kurtosis=3.24748  
 QQ: SampleSize=101 Mean=80.25923 Median=95.85649 stdDev=31.49316 Skewness=-0.99315 Kurtosis=2.51678  
 Reference Line: SampleSize=2 Mean=62.04545 Median=62.04545 stdDev=70.58577 Skewness=0 Kurtosis=0.5  
 Reference Line: SampleSize=2 Mean=55.60662 Median=55.60662 stdDev=75.50964 Skewness=0 Kurtosis=0.5

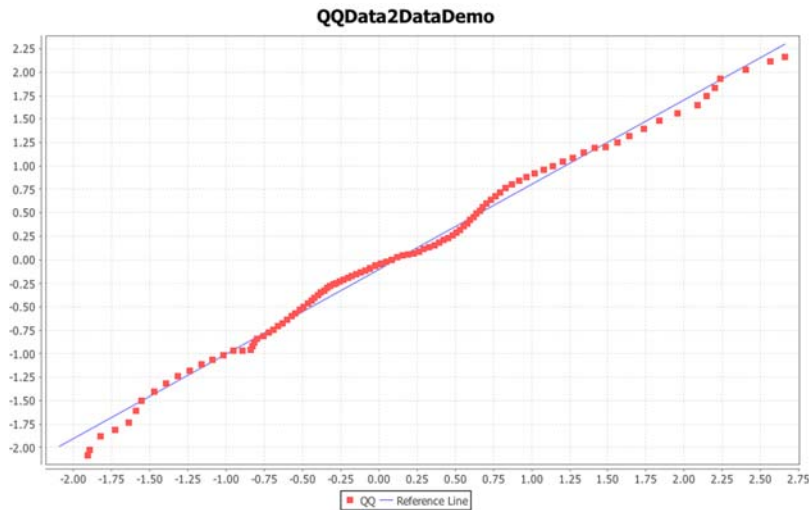
No description file is found for QQData2DataDemo.

Then click the mapping tab and map the R and SOCR column headings into the 2 bins (X and Y axes). Finally click “Update Chart” to plot the QQ plot (quartiles of R and SOCR data are on the X and Y axes, respectively). We can change the Chart Titles and label the axes (using right-click and setting the chart properties), however we did not do that here to avoid possible confusion.

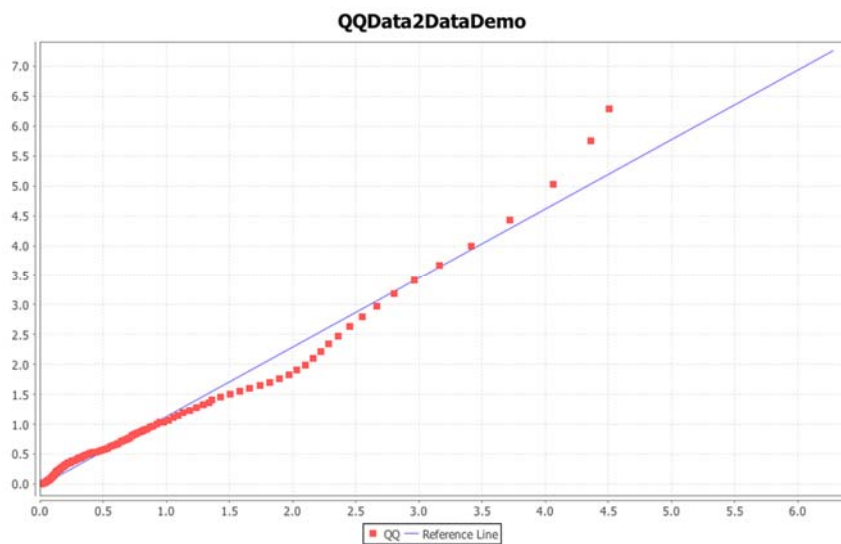
### Uniform QQ Plot



## Normal QQ plot



## Exponential QQ plot



### Notes:

- For SOCR RNG see this activity: [http://wiki.socr.umich.edu/index.php/SOCR\\_EduMaterials\\_Activities\\_RNG](http://wiki.socr.umich.edu/index.php/SOCR_EduMaterials_Activities_RNG)
- For R random number generation:

```
unif.gen <- runif(100, -1, 1)
norm.gen <- rnorm(100, 0, 1)
exp.gen <- rexp(100, 1)
```

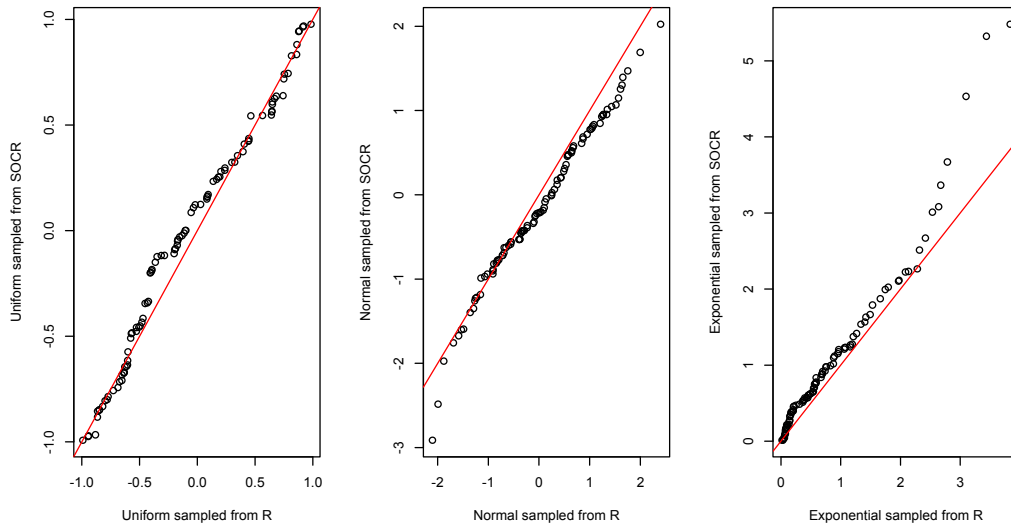
```
dat.gen <- cbind(unif.gen, norm.gen, exp.gen)
write.csv(dat.gen, 'hw_dat_gen.csv')
```

```
# QQ plot in R
data <- read.csv("hw_dat_gen.csv", header=T)
attach(data)
```

```

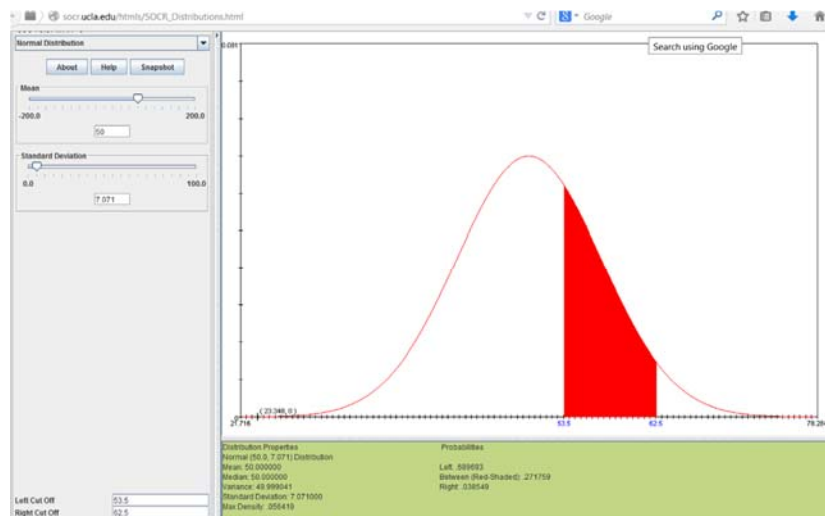
colnames(data)=c('U_R','U_S','N_R','N_S','E_R','E_S')
par(mfrow=c(1,3))
qqplot(U_R,U_S,xlab='Uniform sampled from R',ylab='Uniform sampled from SOCR')
abline(0,1,col='red')
qqplot(N_R,N_S,xlab='Normal sampled from R',ylab='Normal sampled from SOCR')
abline(0,1,col='red')
qqplot(E_R,E_S,xlab='Exponential sampled from R',ylab='Exponential sampled from SOCR')
abline(0,1,col='red')

```



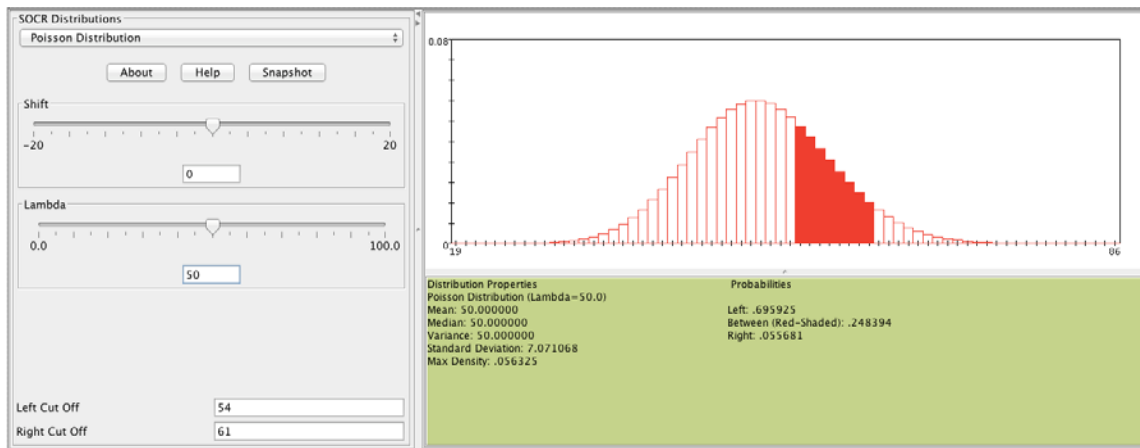
**Problem 3:** The [Poisson distribution with parameter  \$\lambda\$](#)  can be [approximated with normal when  \$\lambda\$  is large](#). Suppose patients arrive at a hospital at a rate of 50 per day. Let's assume that the process is a Poisson random variable with  $\lambda=50$ . Compute the probability that in the next day the number of patients that arrive at this hospital will be between 54 and 62. Use the SOCR [Normal](#) ( $N(\mu=50, \sigma=7.071)$ ) and [Poisson](#) ( $\text{Poisson}(\lambda=50)$ ) distribution calculators.

Using the normal distribution, the probability is  $P(53.5 < X < 62.5 | N(\mu=50, \sigma=7.071)) = \mathbf{0.271759}$  (see the [continuity correction section of the EBook](#)).



Using the Poisson distribution, the probability is  $P(54 \leq X \leq 62 | P(\lambda=50)) = \mathbf{0.248}$ . I do not include 62 because in the continuous normal distribution we only include values up to 62.0

<http://www.socr.umich.edu/people/dinov/2014/Fall/HS851/HWs.html>



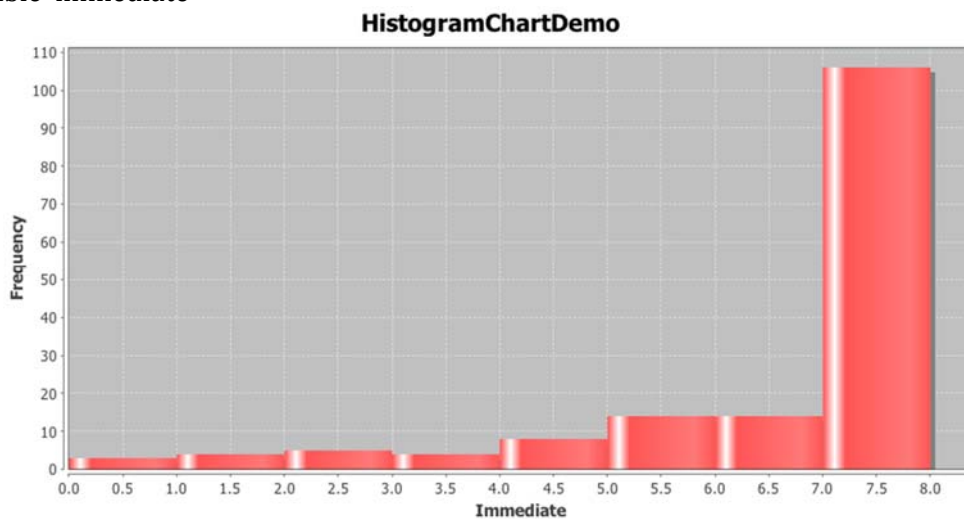
R Code:

```
ppois(q=61, lambda=50, lower.tail=T) -
  ppois(q=53, lambda=50, lower.tail=T)
#I used 61 and 53 to match the results from SOCR.
#I'm not sure what values are being included here...

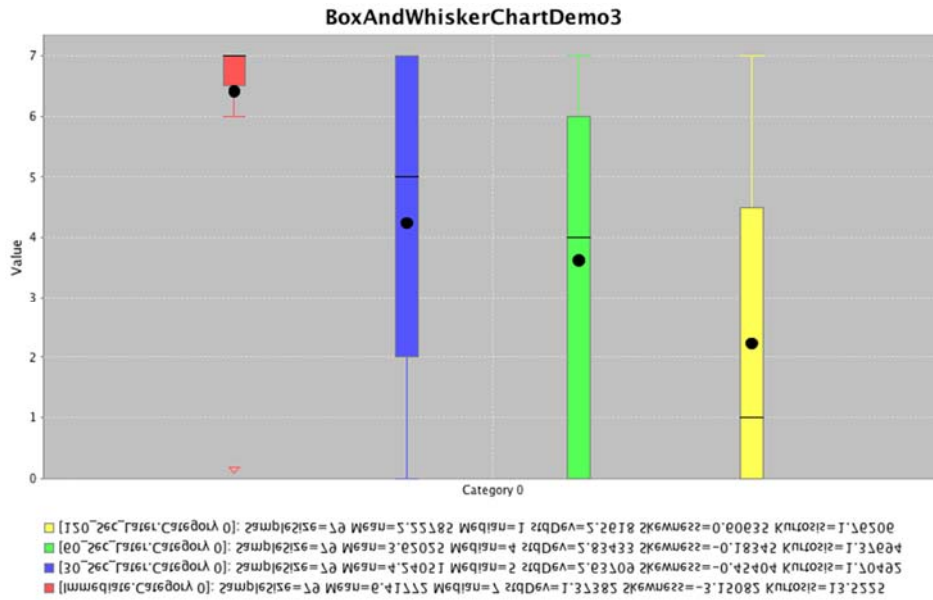
pnorm(q=62, mean=50, sd=7.071, lower.tail=T) -
  pnorm(q=54, mean=50, sd=7.071, lower.tail=T)
```

**Problem 4:** Use the [Vitamin K shots Neonate Infant Pain Score \(NIPS\) dataset](#) and [SOCR Charts](#) to generate a box plot, a scatter plot, a line plot, a dot plot, a boxplot, a histogram plot and a pie chart for some of the variables ([Immediate](#), [30\\_Sec\\_Later](#), [60\\_Sec\\_Later](#), [120\\_Sec\\_Later](#), [Total\\_Cry\\_Time](#)).

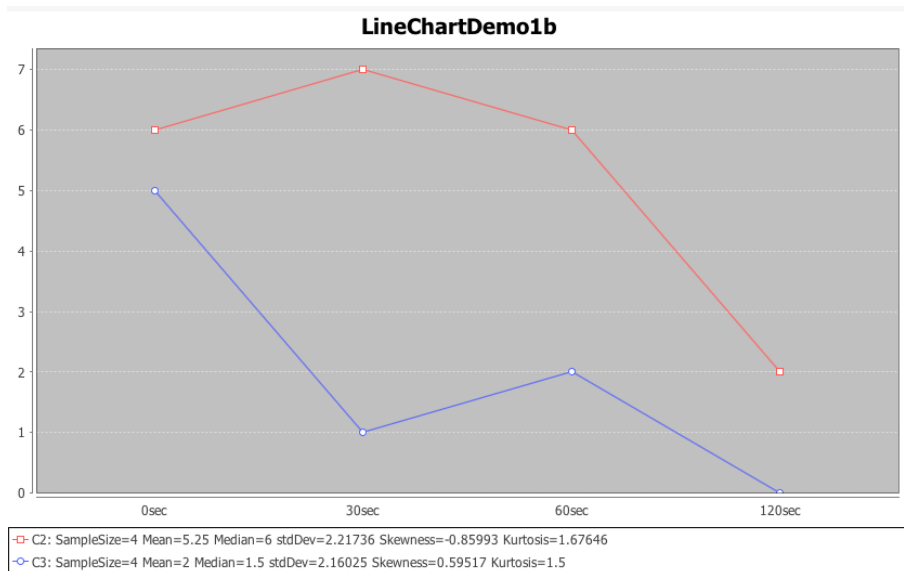
Histogram of variable 'immediate'

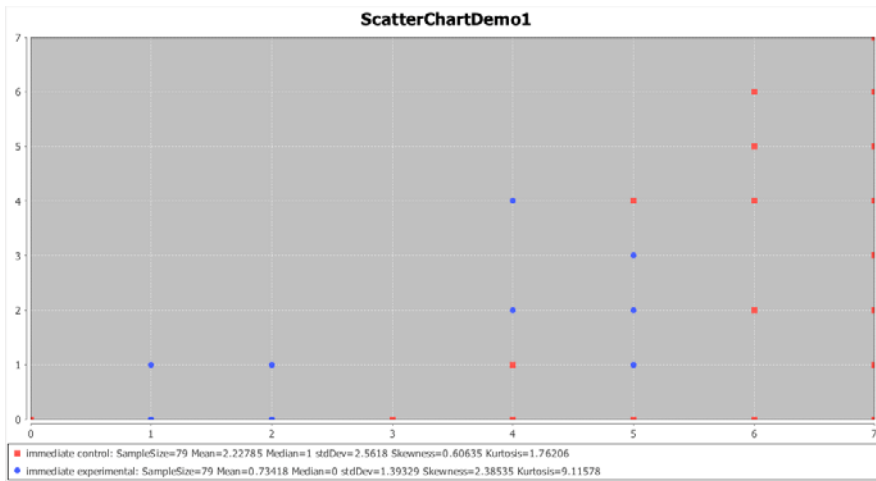


Boxplot of the control group for various variables.

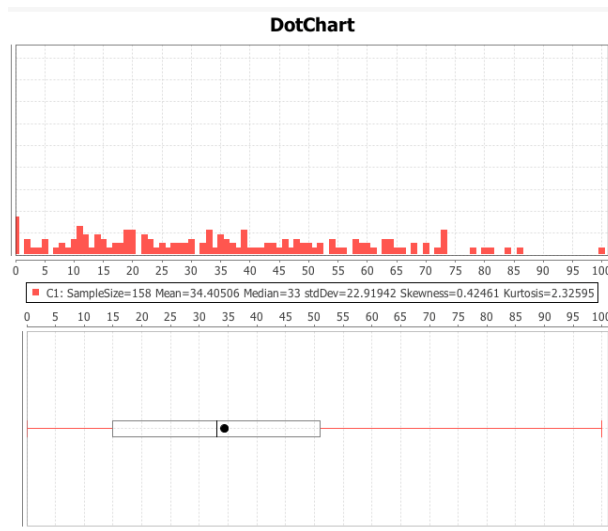


## Line chart

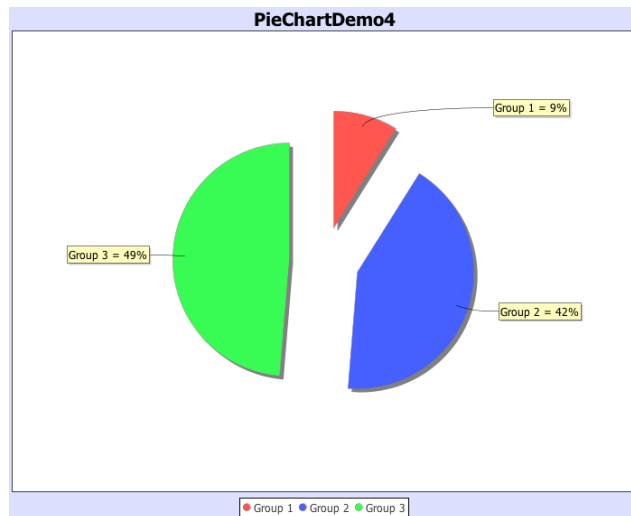




Dot chart of total cry time



Pie chart of clusters



R Code:

```
nips.dat<-read.csv('NIPS_data.csv')
head(nips.dat)
```

<http://www.socr.umich.edu/people/dinov/2014/Fall/HS851/HWs.html>

```
hist(nips.dat$Immediate)
boxplot(nips.dat[nips.dat$Group_NC1_Interv2==1,3:6])
plot(1:4,unlist(nips.dat[1,3:6]), col=1, type='b', ylim=c(0,7))
points(1:4, unlist(nips.dat[2,3:6]), col=2, type='b')

#calculate proportions for pie chart of variable 'cluster'
cluster.tab<-table(nips.dat$Cluster)
cluster.props<-cluster.tab/sum(cluster.tab)
pie(cluster.props, labels=paste('group',names(cluster.props)))
```



**Appendix: Problem 2: Random Number generation results (SOCR and R)**

Index	unif.gen.r	norm.gen.r	exp.gen.r	unif.gen.socr	norm.gen.socr	exp.gen.socr
1	-0.03421	2.146162	1.340007	0.460332	1.20057	1.041762
2	-0.72205	0.629507	1.099317	-0.86515	-0.22798	1.243426
3	-0.67383	-0.66391	0.312262	-0.05886	0.997939	0.589069
4	-0.83861	-0.36714	0.114022	0.654992	1.013191	0.771458
5	-0.12954	-0.31342	0.709416	-0.07565	0.774204	0.493172
6	0.547198	0.591664	0.894984	0.32987	1.547096	3.501474
7	0.20275	-0.72866	0.216729	0.45734	-0.65392	0.292094
8	0.689725	-0.25386	2.546228	-0.71675	0.549533	0.385044
9	0.932319	-0.50089	0.037871	0.449293	0.039298	0.101861
10	-0.23215	-0.37405	2.265916	0.005879	0.155565	0.450133
11	-0.05471	0.71512	0.070696	0.301959	-0.81355	0.633944
12	-0.60256	-0.07206	0.870941	-0.08382	1.665624	0.258207
13	0.309499	0.390936	0.151744	-0.52922	0.221251	4.207319
14	-0.88687	0.151597	2.180663	0.540384	-0.2561	2.889952
15	-0.70833	-0.35227	0.200772	-0.51099	-0.34571	0.084309
16	-0.55618	-1.27415	0.738722	0.734095	0.691379	0.510943
17	0.997966	-0.37241	0.723516	0.108784	-0.07839	2.84049
18	-0.31476	-0.74948	0.725648	0.753113	-0.21449	0.938006
19	0.648257	-1.48634	2.746657	0.399	-0.93579	0.872644
20	-0.2527	1.636721	0.711131	-0.53996	-1.57178	0.005215
21	0.488616	-0.23512	3.036433	-0.60762	0.274927	0.845941
22	-0.62573	0.613624	1.974556	0.229866	0.916813	5.327805
23	0.546316	0.092615	0.105033	-0.77381	-0.70757	0.466315
24	-0.17407	-0.43698	0.578787	-0.32124	-0.18651	1.501219
25	-0.3447	-1.88449	0.270065	0.737515	-1.6356	0.603368
26	0.308256	-0.73031	3.337005	-0.04694	-0.06189	1.905998
27	-0.26303	0.663724	0.219036	0.750205	0.232279	0.186334
28	0.531543	1.71799	0.407216	0.529095	0.073917	3.151731
29	0.602814	0.717651	4.511306	0.453735	1.249572	1.568328
30	0.978829	0.293492	0.114425	0.118894	-2.02146	0.081836
31	-0.85019	-1.4763	1.124781	-0.51768	-0.28394	0.340263
32	0.23253	1.475606	0.599163	0.963082	-2.97734	1.290559
33	-0.28099	1.494191	1.223023	0.745283	0.44355	0.923142
34	-0.30193	-0.28436	0.700383	-0.5863	0.064641	1.869083
35	0.548397	-0.20229	0.086656	-0.54248	0.093079	6.278972
36	0.995536	-0.70494	0.285405	0.939171	-0.40414	0.447815
37	-0.01905	-0.17924	0.071937	-0.07612	-0.31802	0.937193
38	-0.96642	-0.68723	2.111835	-0.523	2.167942	0.843612
39	0.107123	0.545174	0.974293	0.726575	0.941747	0.08351
40	-0.06	0.790998	0.68405	0.153013	-0.17526	0.39209
41	-0.70203	1.298077	0.969277	0.365026	-1.14691	0.034582
42	-0.4387	-0.21428	3.832193	-0.20345	0.156672	0.315035
43	-0.45316	0.332357	4.407625	0.948052	-1.34833	0.131123

Index	unif.gen.r	norm.gen.r	exp.gen.r	unif.gen.socr	norm.gen.socr	exp.gen.socr
44	0.655648	-1.86569	1.429922	0.208282	-0.28243	0.153636
45	0.3256	-1.34245	0.843509	0.116541	-0.25856	0.881637
46	0.983889	2.12348	0.857802	-0.13949	-0.85751	1.545764
47	0.300022	-1.83606	2.056815	-0.61146	-2.41567	0.071218
48	-0.73311	-0.38427	0.09266	0.491245	-0.81568	0.728199
49	0.509772	0.748845	2.212804	-0.99243	1.052579	3.929189
50	0.173622	2.657411	0.0562	-0.9653	-0.82528	0.968741
51	-0.76479	-2.28273	2.791011	0.109822	-0.27364	0.01077
52	-0.22208	0.472962	0.840846	-0.56335	-0.97879	0.402135
53	0.426459	-1.13678	0.115148	0.608226	-0.57152	0.770245
54	-0.34607	0.567162	3.128469	0.337765	-0.75563	1.20417
55	-0.34911	-0.61365	0.050545	-0.84481	0.649193	0.308639
56	0.927089	0.297586	1.114771	-0.64119	1.228224	0.355358
57	0.743263	-0.41333	2.130998	-0.54559	-0.50897	0.712986
58	-0.78684	-0.10762	1.592749	0.054734	0.528545	0.238807
59	-0.53026	1.182324	0.634011	-0.85744	-0.09481	0.362151
60	-0.75019	-1.67929	0.027767	0.570614	0.077227	1.966582
61	0.441064	-1.34327	0.203299	0.579498	1.803817	0.02477
62	0.115574	0.946397	0.356205	0.959642	-0.77776	1.339688
63	-0.72323	1.002117	0.280054	-0.1695	0.30471	0.229536
64	0.207033	1.174052	0.795948	-0.87549	0.276103	0.744773
65	0.922349	0.312329	0.636844	-0.41622	-1.11926	1.287061
66	0.555111	0.254074	1.743359	0.810238	-1.2959	0.401478
67	-0.21242	0.784066	2.296724	0.943708	1.687488	2.583519
68	-0.58894	-0.05484	0.169041	0.183287	-0.26602	0.368145
69	-0.56253	-0.86096	1.51277	0.224716	0.605233	0.645768
70	0.607497	2.554486	0.274575	-0.26691	0.978547	1.891233
71	-0.30391	0.590389	0.148642	-0.17202	-0.03424	1.640018
72	-0.11632	-2.20915	0.469799	-0.23503	-1.02101	0.052416
73	-0.5117	0.017043	0.212306	0.825396	1.275137	1.560863
74	-0.59448	-1.49431	0.277172	0.134829	-0.16892	1.63781
75	0.644255	-0.96206	0.517558	-0.64993	-1.72645	0.61451
76	-0.36739	-2.30136	0.751569	-0.59528	0.902601	0.564951
77	-0.57243	0.132685	0.772659	0.675575	-0.37315	2.474797
78	-0.42413	1.738395	2.428733	0.855513	-0.57194	0.259538
79	0.216546	1.358673	0.036986	0.866471	0.209621	0.394872
80	0.4865	-0.1048	0.392775	-0.86183	-0.22841	0.19988
81	0.307922	0.466571	0.43465	-0.96086	-0.04075	2.756624
82	0.83845	0.195562	0.024099	-0.57759	2.051262	0.812336
83	0.054557	-0.63174	0.029043	-0.94467	0.444667	0.013513
84	0.991523	-0.02751	0.698759	0.134421	1.801101	0.881565
85	0.237425	0.492225	0.240529	0.383753	0.742141	0.842572
86	0.222587	-0.56496	0.159572	0.467906	0.517282	0.466636
87	0.142296	-0.56339	1.770333	-0.06051	2.049387	1.304635
88	0.025827	-0.4032	1.077106	-0.70687	0.025195	1.042272

Index	unif.gen.r	norm.gen.r	exp.gen.r	unif.gen.socr	norm.gen.socr	exp.gen.socr
89	0.644079	0.820375	0.414582	0.012978	1.379224	1.554755
90	-0.93367	1.666928	0.531248	0.642531	0.378679	2.141496
91	-0.13577	0.570079	0.236922	0.523789	-1.9216	3.945831
92	0.47481	1.215485	2.132891	0.169072	-2.78431	0.476985
93	0.734987	0.964138	0.10621	0.407978	0.897581	1.821299
94	-0.41119	-0.92193	0.888887	0.684376	-0.60674	0.461348
95	0.393878	0.709595	1.286282	0.128438	-1.09462	0.024869
96	0.062104	-0.38311	0.428636	-0.30117	-0.61066	0.893323
97	0.332088	-0.81415	0.12951	0.275516	0.505129	1.127407
98	-0.93	0.624815	1.035004	-0.30138	-1.29674	0.784496
99	-0.33118	2.19848	0.328307	-0.27936	0.052262	0.533495
100	-0.59948	-1.3648	2.060017	0.378161	-1.26381	0.177867