

# rec7

October 29, 2016

## 1 Probability and distributions: uniform distribution

- more interesting than coin flips

```
In [6]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
import matplotlib as mpl
```

```
In [3]: # uniform random number (from [0,1])
np.random.random()
```

```
Out[3]: 0.9840531854509492
```

```
In [5]: # generate a large array of them
randArr = np.random.random((1000,))
print(randArr)
```

```
[ 0.43932662  0.92303056  0.95362779  0.02507379  0.83851937  0.35469688
 0.0320268   0.30406117  0.45883142  0.40152035  0.76160598  0.79198227
 0.27431931  0.760312   0.70432578  0.85127965  0.02784033  0.42482476
 0.2569968   0.29446361  0.22716809  0.52394722  0.14830686  0.66193101
 0.24111876  0.35763802  0.54819984  0.18028755  0.20511505  0.26065344
 0.0129208   0.86549528  0.38462112  0.30245392  0.47046773  0.52114798
 0.64455454  0.23260381  0.53381666  0.06041079  0.68502009  0.57370341
 0.01354409  0.89999626  0.64444706  0.24819795  0.64366703  0.96643648
 0.7465468   0.90491128  0.83019648  0.48611479  0.50787884  0.1105018
 0.48140581  0.35895768  0.8425198   0.51873404  0.78455673  0.58223607
 0.42087894  0.23228898  0.38342084  0.00972764  0.66142659  0.77784459
 0.00397615  0.49816531  0.46996638  0.20360328  0.9385943   0.14654952
 0.14598159  0.89273158  0.15812949  0.5350867   0.97646967  0.82916241
 0.41132515  0.05288579  0.20432874  0.78283273  0.56570872  0.56443003
 0.51594208  0.68050231  0.19661863  0.08043029  0.15016969  0.77658018
 0.97104089  0.58462433  0.74754483  0.6495987   0.10052636  0.02861127
 0.74157117  0.38696799  0.50988199  0.71111949  0.51753063  0.64482443
 0.36589536  0.38969907  0.58907385  0.92638112  0.71451602  0.59843759
 0.06377072  0.35553937  0.47296108  0.24804759  0.66652508  0.3117917
 0.51148843  0.5319085   0.05760238  0.16958877  0.81987293  0.91378895]
```

0.55937975	0.33384374	0.1398821	0.73122633	0.71360137	0.45312736
0.57651532	0.47617282	0.9461809	0.96356311	0.93511668	0.42332374
0.85842456	0.21360256	0.91805317	0.78873065	0.55770878	0.02240738
0.74744714	0.03033391	0.84507015	0.58910723	0.15073373	0.59461225
0.21457496	0.31698386	0.29511025	0.17333893	0.69991064	0.57095528
0.20065638	0.34185228	0.28005629	0.1832709	0.91377323	0.94377108
0.30782698	0.5587544	0.48450338	0.42084325	0.17049464	0.99695954
0.50210858	0.92857921	0.23761042	0.95685219	0.27714764	0.28223261
0.62799588	0.4077952	0.82303217	0.20348169	0.62772484	0.56742936
0.83884385	0.810281	0.15378064	0.50628967	0.15874285	0.99659345
0.50792328	0.67691276	0.18041525	0.69692144	0.50786358	0.88270624
0.59244193	0.52679537	0.15241995	0.16598696	0.35953992	0.16509911
0.9480007	0.08999627	0.92885129	0.15579174	0.37943517	0.07123411
0.38103516	0.22309144	0.26202543	0.59393957	0.23158819	0.44682806
0.69581977	0.75635696	0.06247903	0.18886842	0.92246008	0.76282843
0.5964749	0.5171499	0.06345998	0.73404381	0.16394927	0.36496767
0.1972421	0.53348667	0.67544531	0.41813454	0.96402583	0.58755629
0.34918835	0.70288453	0.09973856	0.86779494	0.85184236	0.10344631
0.95536792	0.90388592	0.11862975	0.51931685	0.96763422	0.87255248
0.13127469	0.77333015	0.81166002	0.66628673	0.80408877	0.21894738
0.0700805	0.60897729	0.64080776	0.58013586	0.85320161	0.56693035
0.8131021	0.63783704	0.25441161	0.59179871	0.54208472	0.77246441
0.75751766	0.71611498	0.93355952	0.69632649	0.03401593	0.84548702
0.46184747	0.05523436	0.1468018	0.31496159	0.47444963	0.25776627
0.55003102	0.404343	0.38498301	0.80882038	0.61959353	0.42884848
0.53195294	0.21601176	0.01105279	0.40119131	0.65309494	0.15388033
0.57859937	0.7321163	0.07577127	0.46993322	0.01524215	0.09393068
0.6961182	0.86144962	0.61262692	0.21246357	0.24680295	0.37929902
0.26945699	0.1713005	0.19412954	0.52561487	0.52870517	0.75511301
0.34056378	0.62820614	0.2273529	0.45282102	0.00702039	0.25784794
0.11116578	0.66115492	0.68624714	0.9439024	0.87152639	0.41086487
0.29715396	0.17541926	0.89771647	0.30535964	0.41981006	0.16236545
0.31062156	0.03667209	0.63214395	0.52224493	0.94954489	0.33884172
0.03649283	0.60126137	0.47659288	0.24433272	0.4372291	0.47979704
0.23857706	0.22216452	0.47182518	0.65444261	0.69217353	0.67754173
0.44212907	0.59419192	0.63549585	0.36383533	0.26357035	0.78064438
0.39462019	0.35581749	0.11699506	0.38069082	0.05322759	0.1466758
0.36499985	0.85950079	0.06211293	0.1845904	0.33616359	0.75150621
0.55483276	0.47152992	0.17451094	0.23515341	0.90755404	0.64995347
0.22558083	0.1929324	0.53486806	0.59087052	0.91899324	0.16706007
0.0456931	0.3863219	0.27136523	0.44569082	0.42172977	0.29638102
0.2127021	0.58231833	0.60884745	0.78896824	0.25065402	0.61507348
0.96483537	0.01722829	0.92264473	0.32004704	0.18459803	0.72114782
0.35852134	0.28698121	0.26868841	0.64706131	0.21180719	0.68089352
0.61090976	0.72536738	0.95704607	0.42853144	0.13518385	0.37613695
0.43585021	0.99413596	0.81712674	0.63559922	0.72458942	0.66116028
0.37253127	0.74794058	0.78464325	0.57293709	0.4576471	0.58393181
0.54138305	0.21701105	0.33881734	0.48459282	0.37691413	0.88644046

0.70501777	0.51802461	0.12541935	0.37590121	0.38558576	0.07252171
0.64777001	0.69315754	0.46176754	0.8868352	0.10376306	0.59170357
0.96743571	0.94043899	0.17716187	0.84060132	0.78314741	0.73924145
0.59247856	0.12220017	0.13563473	0.20155236	0.40447085	0.44215249
0.53658476	0.37558228	0.25910679	0.6471773	0.02491509	0.63812034
0.73069278	0.51927649	0.79606376	0.41565004	0.38767869	0.81450939
0.02341175	0.20117781	0.82433294	0.19301367	0.88930985	0.71171978
0.96106304	0.64866565	0.67235651	0.61693143	0.26005517	0.09376203
0.84433942	0.93984483	0.83656237	0.78064304	0.63337271	0.45928152
0.38751981	0.03315154	0.05642606	0.2801715	0.9399134	0.08353187
0.75063835	0.36924765	0.41062081	0.83278318	0.08746208	0.74671898
0.21883988	0.06179449	0.53010285	0.17432853	0.01595978	0.02526149
0.9774807	0.76380354	0.23489742	0.3417291	0.49217474	0.62758612
0.71593555	0.6909017	0.40052395	0.89378405	0.30500384	0.00107731
0.62565726	0.56515028	0.31893145	0.86382201	0.54015112	0.37985083
0.64833083	0.55625214	0.77958188	0.26198799	0.94677457	0.0416255
0.20750805	0.01597452	0.27092037	0.94590368	0.84403839	0.51537274
0.2279445	0.04948043	0.27373463	0.74225624	0.11514565	0.44441123
0.58837288	0.57407521	0.92857728	0.32328976	0.24894367	0.71712255
0.94981168	0.89542237	0.57874362	0.78587004	0.79987355	0.15739754
0.49298998	0.24889945	0.05244232	0.22076941	0.45194953	0.9804843
0.14452743	0.89307949	0.98562131	0.63688136	0.47573909	0.27929003
0.69467361	0.51792978	0.55315097	0.74848226	0.49749962	0.55826048
0.93208831	0.57708435	0.79198876	0.33811463	0.10287798	0.07309108
0.78789854	0.35694121	0.53000223	0.95328659	0.26303009	0.06476976
0.71839579	0.36298311	0.22415733	0.696899	0.00560628	0.42091965
0.4822852	0.23493967	0.46197674	0.07879657	0.37527484	0.4229862
0.74152402	0.41080962	0.31327701	0.99964755	0.55470863	0.76897624
0.31697359	0.56788991	0.06707363	0.36922984	0.15903509	0.25426496
0.28542875	0.13175239	0.13635778	0.08645855	0.38159811	0.35913535
0.86978696	0.60406654	0.30695474	0.13328814	0.61185743	0.68056821
0.60225342	0.65218193	0.52785508	0.76761758	0.60226084	0.40665953
0.87791654	0.32112559	0.88030398	0.78836286	0.31238076	0.57444553
0.11706464	0.44433153	0.67158736	0.11295846	0.66481026	0.23518386
0.62025794	0.62703058	0.40861567	0.03405474	0.73392949	0.83897385
0.40101056	0.15268419	0.05176716	0.07146226	0.08567706	0.62121937
0.1371628	0.37589467	0.6339966	0.12333397	0.16079589	0.00688329
0.98972781	0.86112564	0.89026177	0.91667138	0.2114442	0.06871565
0.81960936	0.15094872	0.23652675	0.75322694	0.79731321	0.31677258
0.44228134	0.23089302	0.02787079	0.23746463	0.68696909	0.71704725
0.743084	0.08254355	0.79241376	0.13948386	0.61654134	0.82328126
0.52182727	0.56237359	0.76896202	0.43471569	0.13894323	0.80243298
0.71944349	0.10209351	0.51426997	0.75030891	0.71225452	0.16234182
0.53910865	0.59798141	0.078647	0.71054931	0.23211762	0.45758345
0.59589025	0.43303924	0.26491673	0.47411696	0.45742957	0.58619684
0.36556212	0.15285757	0.82944139	0.59493611	0.45095902	0.87745856
0.38340057	0.69823855	0.78067903	0.942645	0.1202483	0.52081195
0.48249701	0.15176944	0.40545672	0.35689481	0.68621365	0.60559628

0.76421267	0.60606102	0.03110383	0.66821167	0.18707543	0.92368288
0.76145741	0.07585238	0.24282336	0.04216631	0.00382804	0.12557041
0.45153597	0.57625067	0.68324615	0.02624207	0.68814164	0.78145755
0.46624608	0.18040665	0.62592323	0.88037118	0.66442765	0.96668996
0.63013417	0.25797788	0.40656162	0.29106235	0.66247345	0.48487383
0.34088786	0.72081451	0.40848089	0.2826707	0.77383892	0.33494538
0.71879195	0.29774766	0.93991696	0.37779703	0.30689075	0.76684364
0.71762957	0.32355876	0.24894709	0.97990419	0.14187615	0.92538658
0.3940143	0.68573711	0.21937102	0.06815661	0.00903256	0.39850999
0.48630651	0.28022485	0.6103149	0.15803198	0.88045344	0.59132543
0.70533901	0.41980659	0.63125881	0.61234076	0.0386069	0.84003354
0.27399834	0.59956	0.01123798	0.0180974	0.30245131	0.01319276
0.96865767	0.4870637	0.53520469	0.4870407	0.3477795	0.54467148
0.59668636	0.53209259	0.54156836	0.56765144	0.84950404	0.5546395
0.89534344	0.79172727	0.3048083	0.33351259	0.15387158	0.40451773
0.88139497	0.40954657	0.84437798	0.79089875	0.50192308	0.72365858
0.43961813	0.83401361	0.69087981	0.64755321	0.42454302	0.49014904
0.80570654	0.52215229	0.74138136	0.51970188	0.94015521	0.45323925
0.27184695	0.45838716	0.32975391	0.67466557	0.44784556	0.87668905
0.75633446	0.15328136	0.70983551	0.10374459	0.20928868	0.3356614
0.02173325	0.78965579	0.36034044	0.94537085	0.87711555	0.2168508
0.16440045	0.7011963	0.98962202	0.7677211	0.85081785	0.46308655
0.90030043	0.31634549	0.38279634	0.86585344	0.96900432	0.74160708
0.24669413	0.51802949	0.89812708	0.06815765	0.46831067	0.73316151
0.93552231	0.86284065	0.27129189	0.97798991	0.58013436	0.38920452
0.12002624	0.05909501	0.4710373	0.73818626	0.06822255	0.81306965
0.38923105	0.92115704	0.3267842	0.69148187	0.0044167	0.25032246
0.37856981	0.18203591	0.44455971	0.08204781	0.57472614	0.77782608
0.2618245	0.04247161	0.65407639	0.48763423	0.27769929	0.3143389
0.32886905	0.25169513	0.13663463	0.15443939	0.83095545	0.45081846
0.82557636	0.73358935	0.79824905	0.60785886	0.85496023	0.48079173
0.79725715	0.72354572	0.11043345	0.37616595	0.21133766	0.53683559
0.80452218	0.66316972	0.23164575	0.29154686	0.77443356	0.72173494
0.69754316	0.29872704	0.19668142	0.50399012	0.57454168	0.97218849
0.4767616	0.10520071	0.19278107	0.59386833	0.93614735	0.99428087
0.64618155	0.46946538	0.81803433	0.37512709	0.94965987	0.22297383
0.85204064	0.15867521	0.34016733	0.62627007	0.61050765	0.16073221
0.93774393	0.73099545	0.08146973	0.97811464	0.18508904	0.50458728
0.81031978	0.58225243	0.21309503	0.41624688	0.27090479	0.38395516
0.88318522	0.41388706	0.07608012	0.78417215	0.82654159	0.31484083
0.2705083	0.397687	0.71524916	0.10315321	0.62014949	0.04299589
0.95053869	0.87185128	0.39792691	0.10222794	0.49194632	0.54197526
0.63745922	0.07116787	0.16742693	0.33192765	0.68103499	0.61395043
0.17392526	0.89001684	0.95276996	0.01353423	0.96762221	0.63234818
0.44633108	0.07266725	0.0280256	0.98085451	0.3969086	0.36787442
0.70924552	0.21341708	0.78733988	0.55147155	0.00175816	0.6206519
0.32630949	0.22801766	0.54388213	0.65927125	0.34628331	0.8289732
0.94455641	0.66675733	0.72155317	0.75097113	0.14694075	0.82850618

```
0.93272662 0.89303781 0.79328387 0.29530145 0.52887442 0.86150836
0.27020302 0.6402873 0.05707078 0.50577085 0.53507581 0.65536856
0.17764123 0.9701673 0.52093017 0.37186804]
```

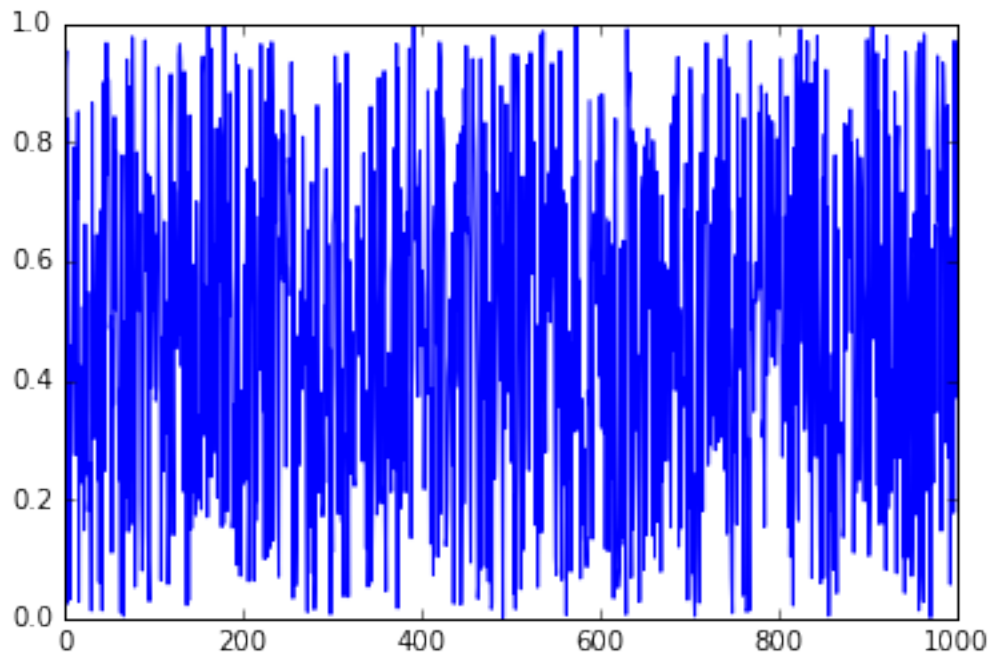
```
In [11]: np.set_printoptions(threshold=20)
         print(randArr)
```

```
[ 0.43932662  0.92303056  0.95362779 ...,  0.9701673  0.52093017
 0.37186804]
```

## 1.1 Use plots to look at it

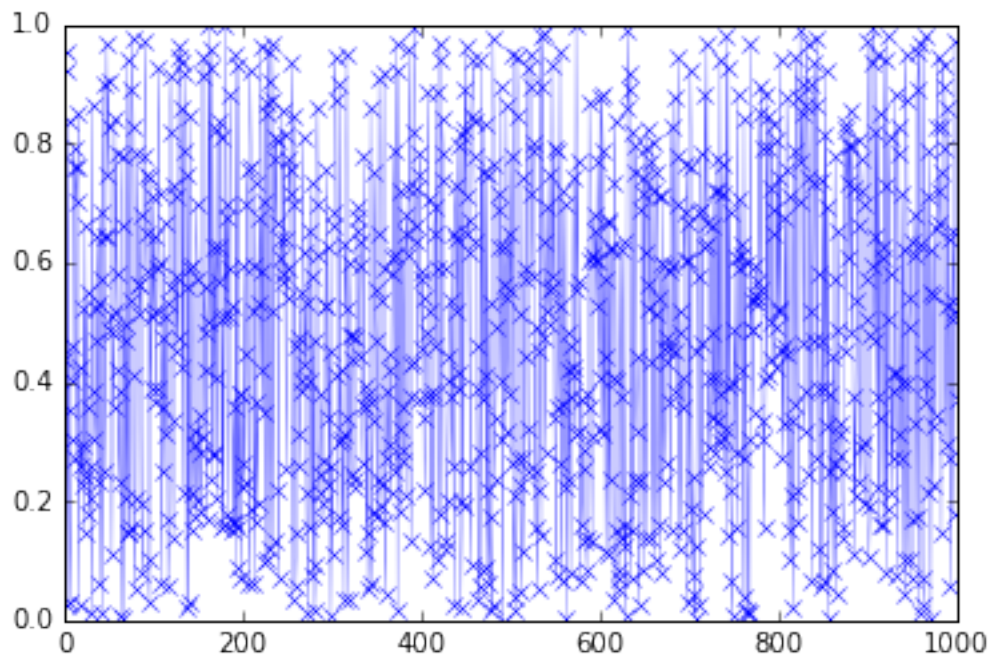
```
In [12]: plt.plot(randArr)
```

```
Out[12]: [<matplotlib.lines.Line2D at 0x10cbadd8>]
```



```
In [17]: plt.plot(randArr, marker='x', linewidth=0.2)
```

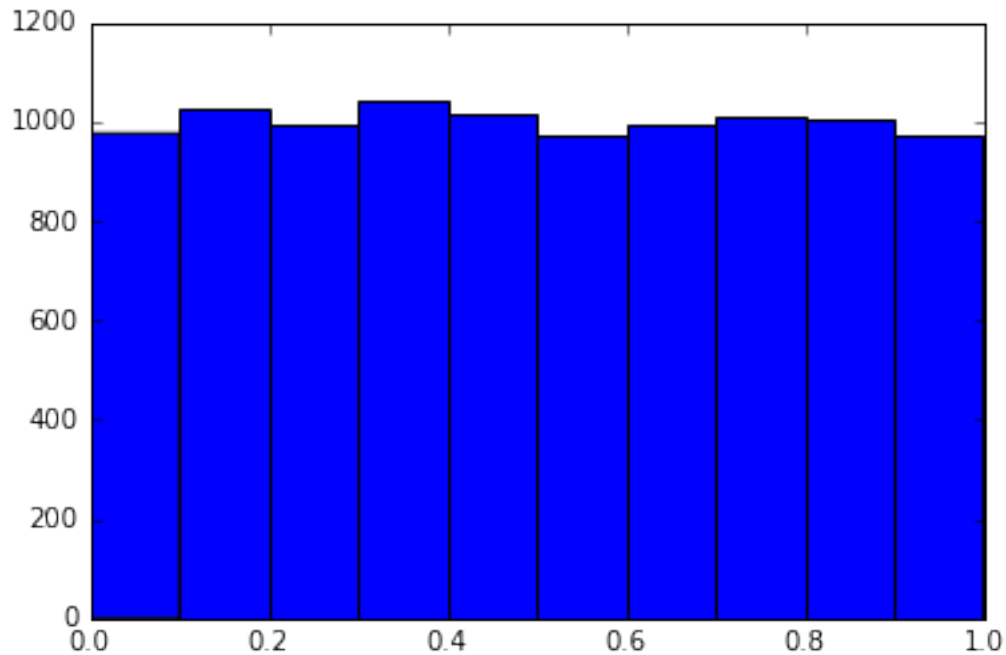
```
Out[17]: [<matplotlib.lines.Line2D at 0x10dd91320>]
```



## 1.2 Histogram

```
In [19]: bigArr = np.random.rand(10000,)
plt.hist(bigArr)
plt.xlabel('value')
plt.ylabel('number in bin')
plt.title('histogram')
```

```
Out[19]: (array([ 976., 1024., 996., 1042., 1013., 971., 991., 1009.,
                  1004., 974.]),
          array([ 1.97123596e-04, 1.00176522e-01, 2.00155920e-01,
                  3.00135319e-01, 4.00114717e-01, 5.00094115e-01,
                  6.00073514e-01, 7.00052912e-01, 8.00032310e-01,
                  9.00011709e-01, 9.99991107e-01]),
          <a list of 10 Patch objects>)
```



```
In [35]: # mean
         np.mean(bigArr)
         # measures of spread
         #np.var(bigArr)
         #np.std(bigArr)
```

```
Out[35]: 0.49880681536795812
```

### 1.3 SEM and confidence intervals

```
In [34]: nPts = len(bigArr)
         sem = np.std(bigArr)/np.sqrt(nPts)
         print(nPts)
         print(sem)
```

```
10000
0.00287522890105
```

```
In [29]: # no equation, so we simulate the 95% CI for the mean
         nReps = 1000
         nPts = 1000
         meanV = np.zeros((nReps,))*np.nan
         # [show meanV here]
         for iR in range(nReps):
             tArr = np.random.random((nPts,))
```

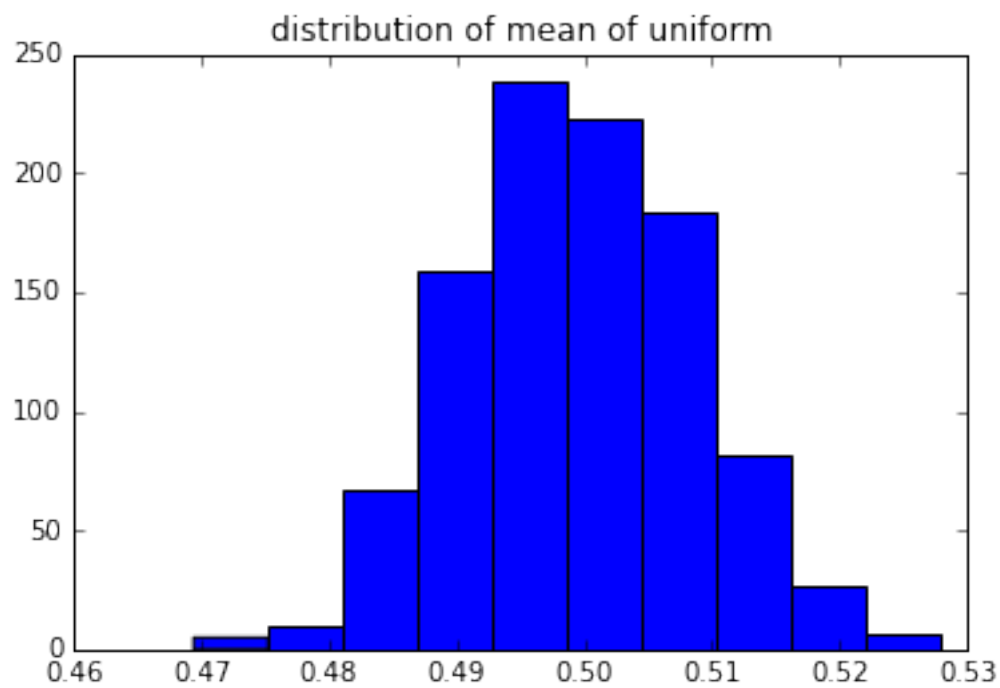
```
meanV[iR] = np.mean(tArr)
```

```
print(np.percentile(meanV, [2.5, 97.5]))
```

```
[ 0.48270442  0.5174354 ]
```

```
In [39]: plt.hist(meanV)
         plt.title('distribution of mean of uniform')
```

```
Out[39]: <matplotlib.text.Text at 0x10ee46c50>
```



---

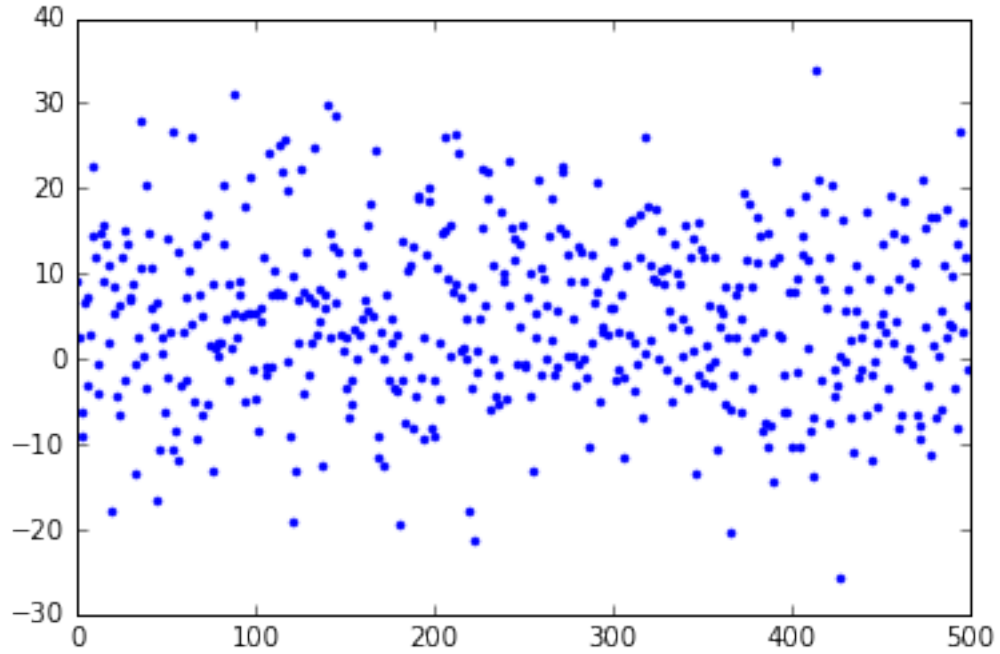
## 2 Normal dist

```
In [53]: #samples
         normArr = np.random.normal(loc=5.0, scale=10.0, size=(1000,))
```

```
In [50]: plt.plot(normArr[0:500], '.')
```

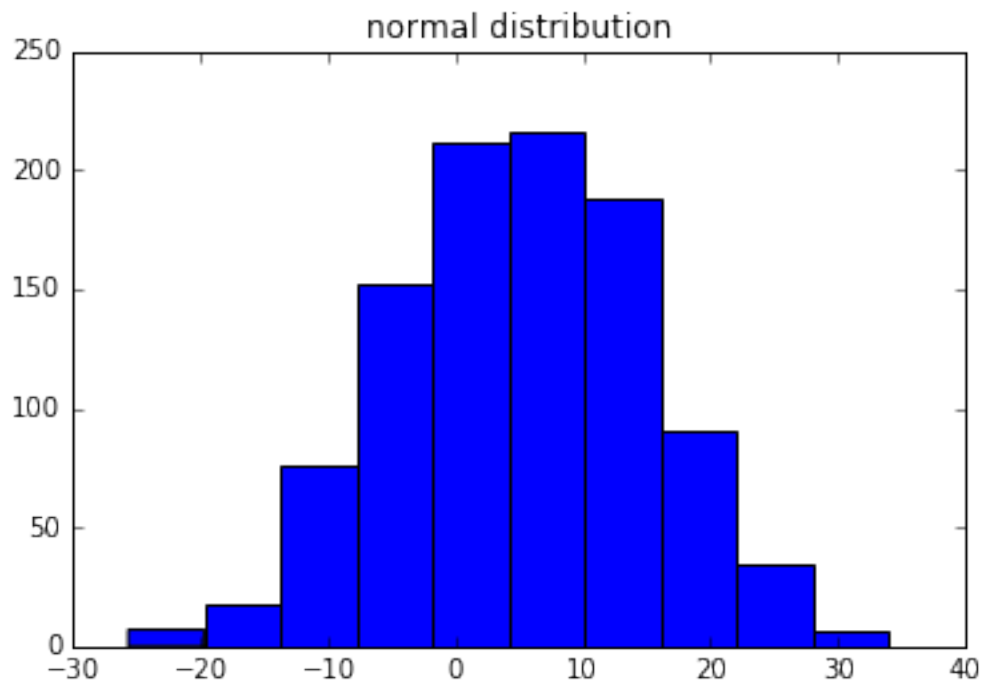


```
Out[50]: [<matplotlib.lines.Line2D at 0x10efd4400>]
```



```
In [51]: plt.hist(normArr)
         plt.title('normal distribution')
```

```
Out[51]: <matplotlib.text.Text at 0x10f07a390>
```



```
In [56]: mu = np.mean(normArr)
         print(mu)
```

```
4.84441322106
```

```
In [57]: np.std(normArr)
```

```
Out[57]: 9.8207158908412904
```

```
In [55]: sem = np.std(normArr)/np.sqrt(len(normArr))
         print(sem)
```

```
0.310558304685
```

**Equation for the 95% CI from a SEM, assuming data are normally distributed**

$$CI_{95} = mean \pm 1.96 * SEM$$

```
In [56]: CI_95 = (mu - 1.96*sem, mu + 1.96*sem)
         print(CI_95)
```

```
(4.3985522253596905, 5.6159407797236369)
```

```
In [ ]: # simulate and see if we get it right
        nReps = 1000
        nPts = 1000
        meanV = np.zeros((nReps,))*np.nan
        # meanV here?
        for iR in range(nReps):
            tArr = np.random.normal(5, 10, (nPts,))
            meanV[iR] = np.mean(tArr)

        print(np.percentile(meanV, [2.5, 97.5]))
```

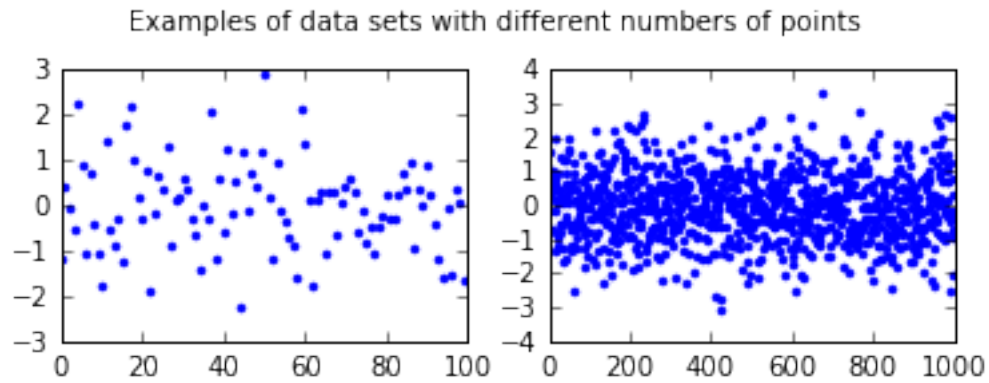
## 2.0.1 SEM as a function of number of trials

```
In [73]: data1 = np.random.normal(0,1,100)
         data2 = np.random.normal(0,1,1000)

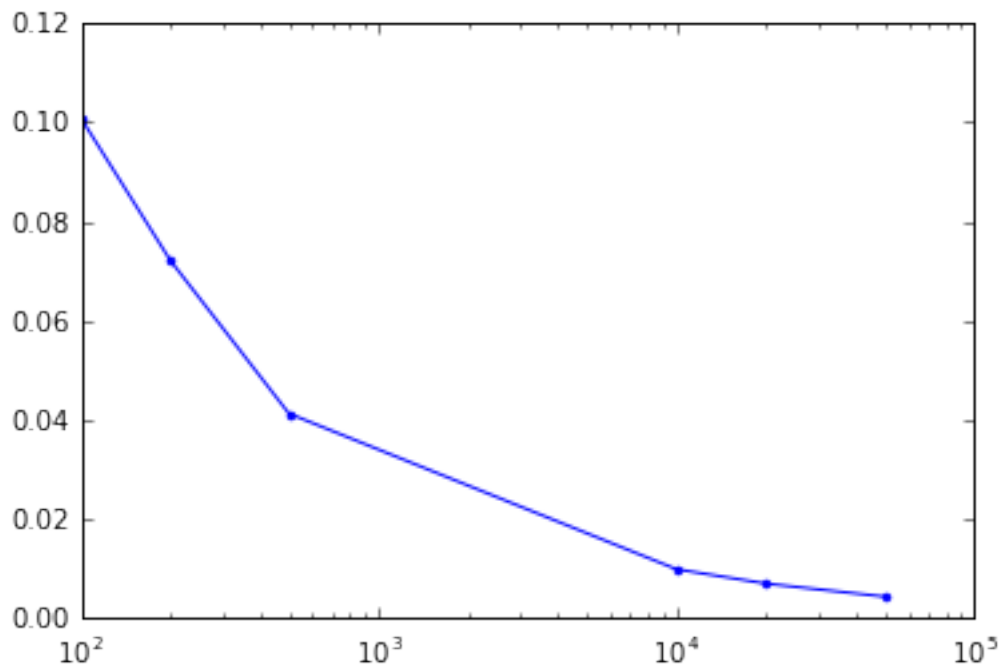
         gs = mpl.gridspec.GridSpec(2,2)
         figH = plt.figure()
         axH = plt.subplot(gs[0,0])
```

```
plt.plot(data1, '.')
axH = plt.subplot(gs[0,1])
plt.plot(data2, '.')
plt.suptitle('Examples of data sets with different numbers of points')
```

Out[73]: <matplotlib.text.Text at 0x11aacfd68>

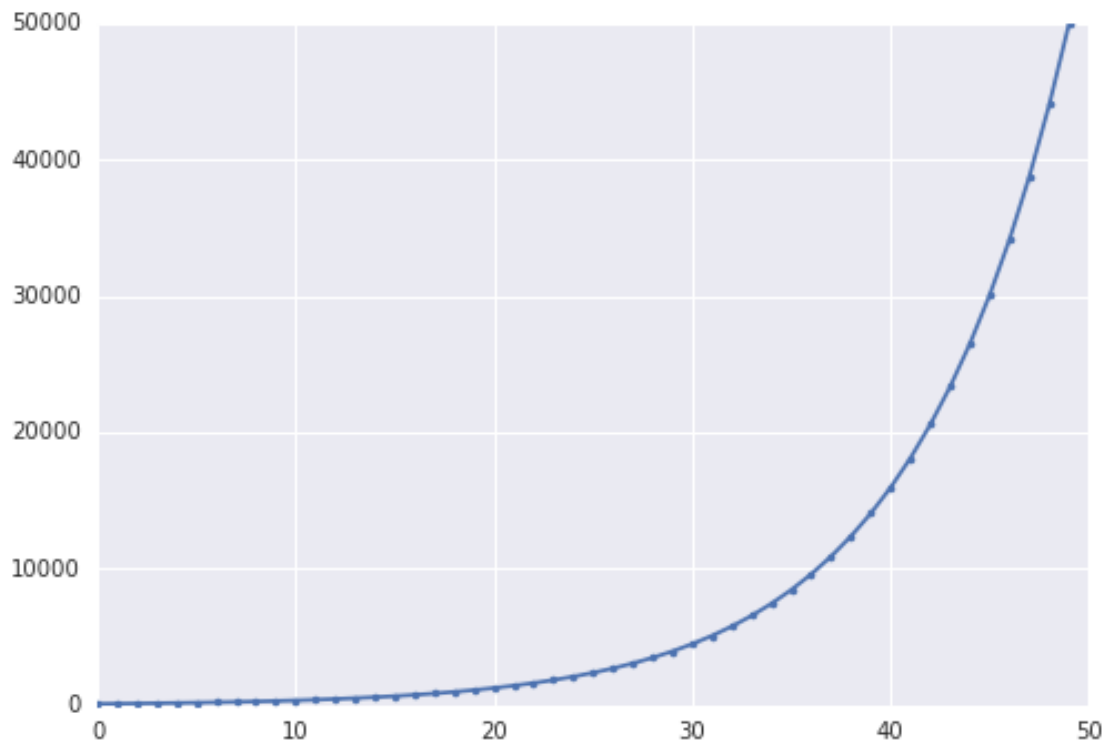


```
In [81]: #xvals = [100,200,500,10000,20000,50000]
semV = np.zeros(len(xvals))*np.nan
for (iX,tX) in enumerate(xvals):
    semV[iX] = np.std(np.random.normal(0,1,[tX,]))/np.sqrt(tX)
#plt.plot(semV, '-.')
plt.plot(xvals, semV, '-.')
axH = plt.gca()
axH.set_xscale('log')
```



```
In [86]: xvals = [100,200,500,10000,20000,50000]
xvals = np.logspace(np.log10(100),np.log10(50000))
plt.plot(xvals, '-')
```

```
Out[86]: [<matplotlib.lines.Line2D at 0x115d150f0>]
```



```
In [83]: import seaborn as sns
```

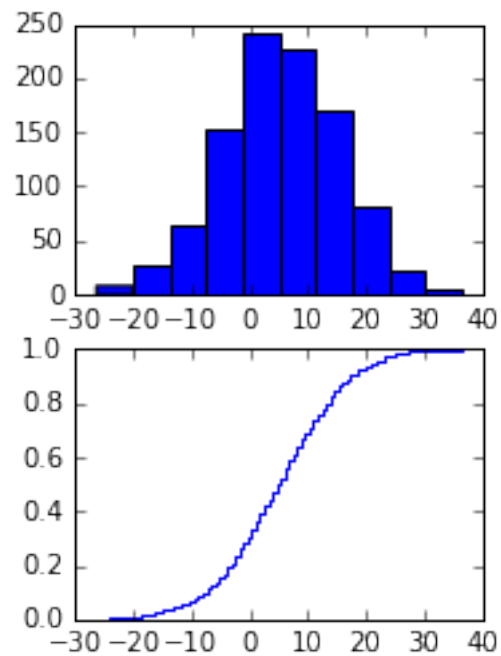
## 2.1 Hist vs CDF

```
In [68]: import statsmodels.api as sm # recommended according to the docs

gs = mpl.gridspec.GridSpec(2,2)
figH = plt.figure()
axH = plt.subplot(gs[0,0])
plt.hist(tArr)
xmin,xmax = axH.get_xlim()

axH = plt.subplot(gs[1,0])
xvals = np.linspace(xmin, xmax, 100)
p = sm.distributions.ECDF(tArr)(xvals)
plt.step(xvals,p)
```

Out[68]: [<matplotlib.lines.Line2D at 0x119eb2278>]



## 2.2 Probability (combinatorics)

- Fair coin:
- prob of heads, tails?
- Biased coin: heads 0.2, tails?
- probability of 2 tails in a row?
- probability of 10 tails in a row?
- 4 sided die:
- probability of rolling a 1
- probability of rolling sum 3