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Assignment: DS298 Assignment1

Problem 2:

Given that X is a random variable representing the skill level of the player. We start by transforming the probability density of X into the probability density of S using the relation $S = (1 + X)^{kX}$. To do this, we use the change of variable method. The density of S , $f_S(s)$, can be expressed as:

$$f_S(s) = f_X(x) \times \left| \frac{dx}{ds} \right|$$

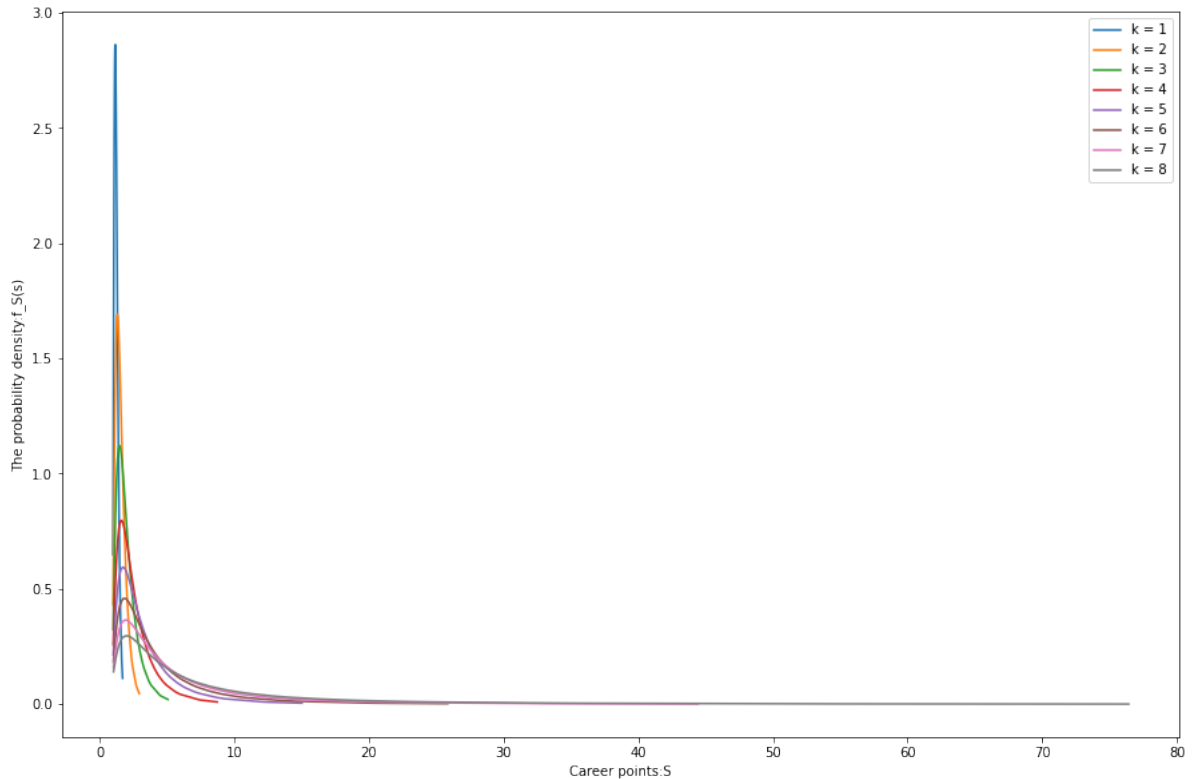
where x is a function of s and can be found by solving the equation $s = (1 + x)^{kx}$ for x . Taking the derivative of s with respect to x , we get:

$$\frac{ds}{dx} = kx(1 + x)^{(kx-1)} + (1 + x)^{kx} \times \ln(1 + x)$$

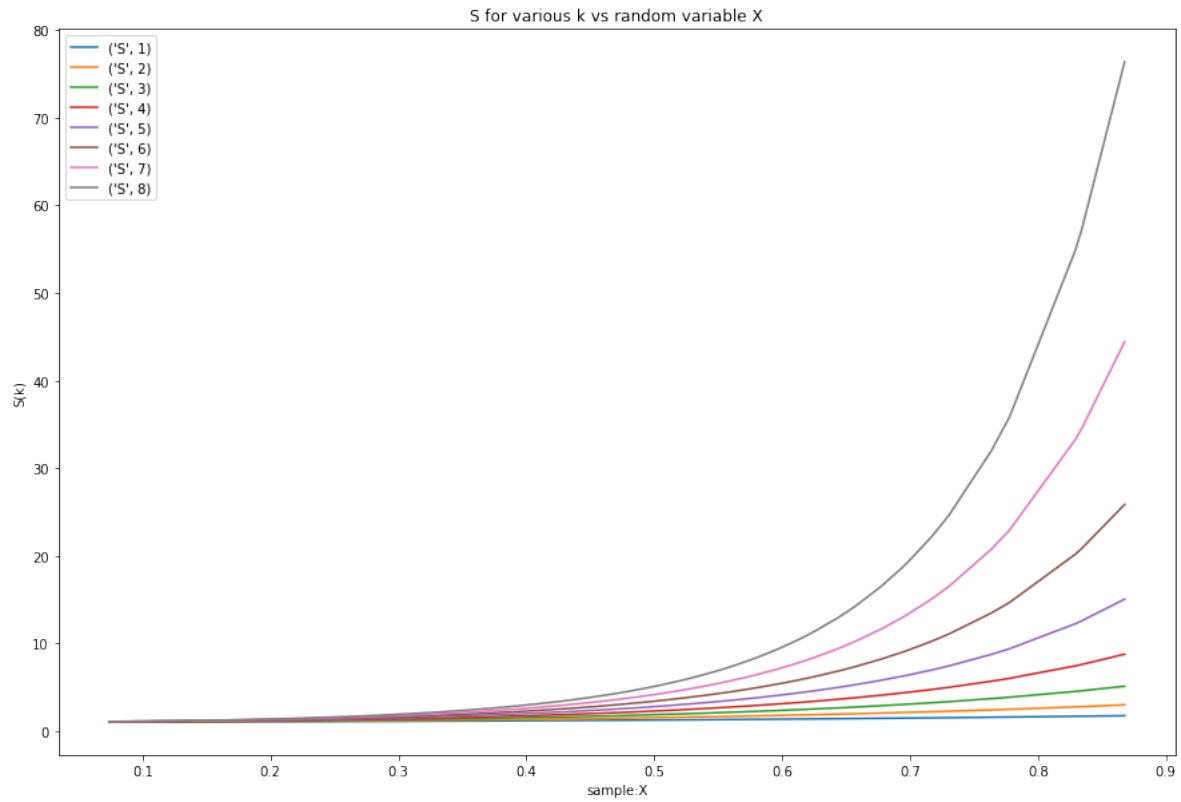
Then we Substitute this into the formula for $f_S(s)$.

After finding x as a function of s , the density of X , $f_X(x)$, can be evaluated at x for each value of s . Finally, we plot $f_S(s)$ as a function of s for each value of $k = 1, 2, \dots, 8$.

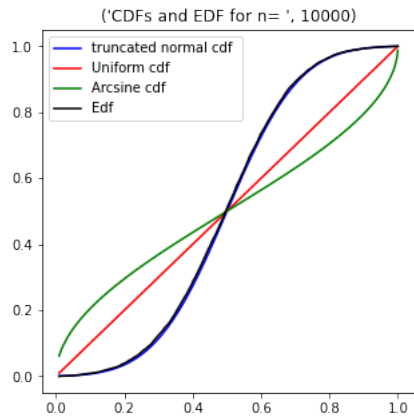
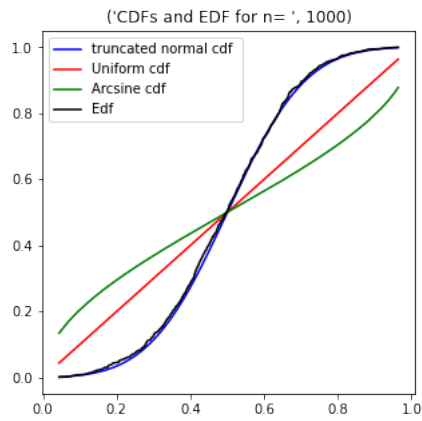
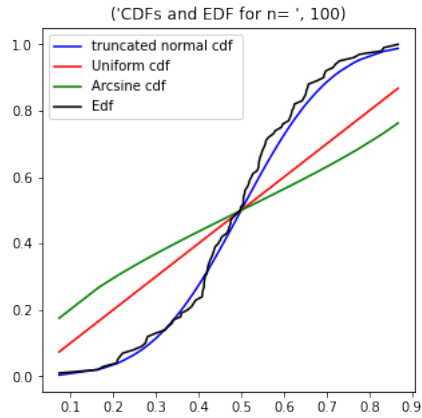
The resulting plot shows the distribution of career points S after each tournament for $k = 1, 2, \dots, 8$. As we can see, the distribution of S becomes more peaked and shifted to the right as the number of tournaments increases, indicating that players who perform well in the early tournaments have a higher probability of earning more career points.

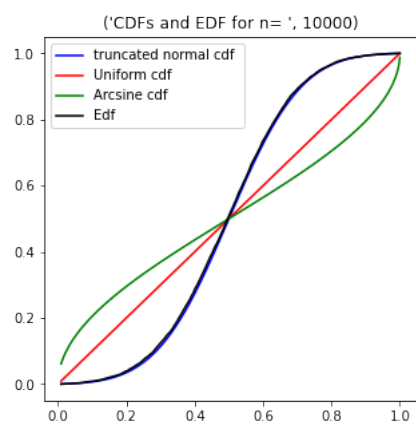
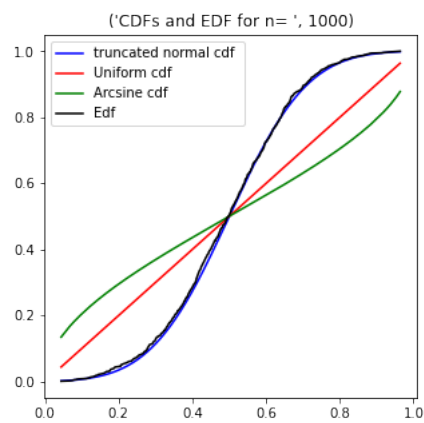
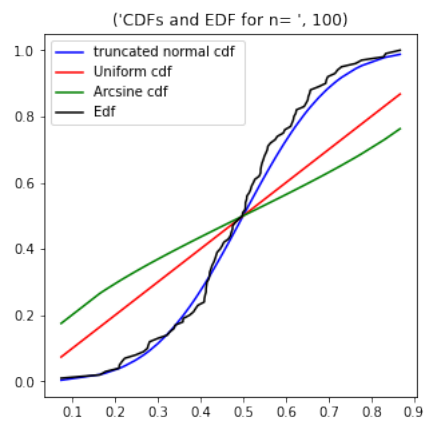
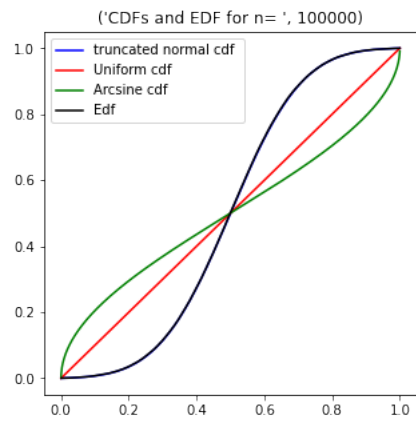


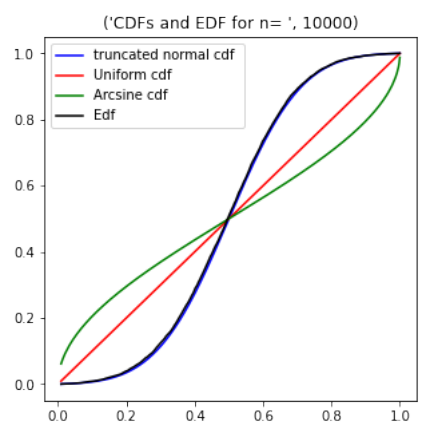
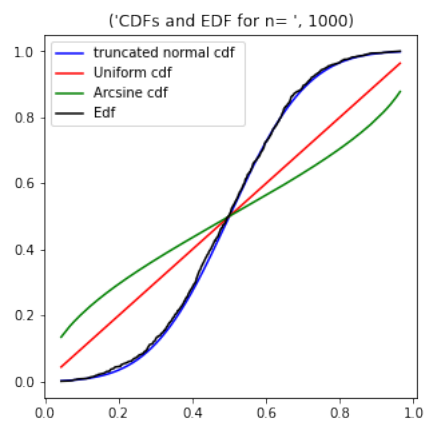
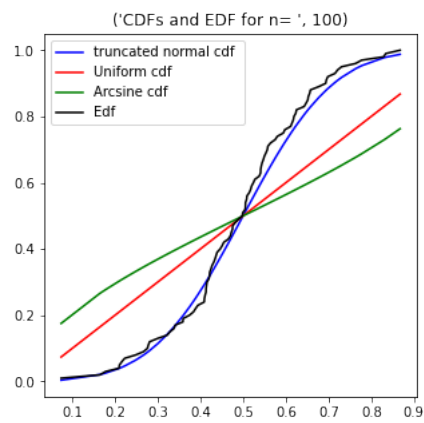
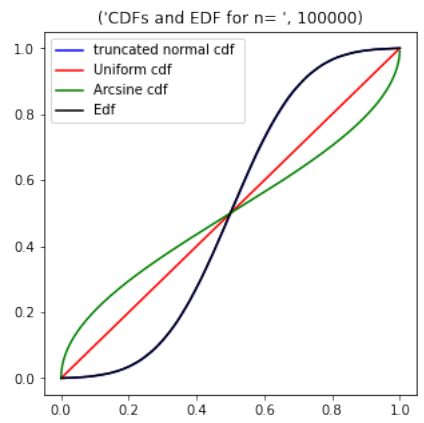
Now we see that career point with the different values of k increases when skill level of player increases. initially for $k = 1$ career point almost constant for different skill level .But for $k = 8$ the career point increases exponentially with increasing value of skill level (X).

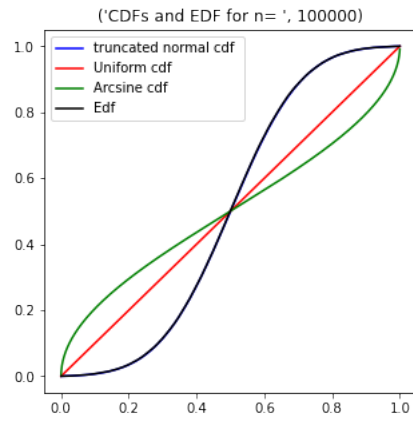


Problem1: Samples X are drawn from Truncated Normal Distribution $N(1/2, 1/36)$ and EDF is plotted along with CDFs corresponding to X assuming it to be drawn from Truncated Normal, Uniform and Arcsine distribution. Here we tested for 3 trials, and see that the EDF (empirical distribution function) nearly close to Truncated Normal CDF. Hence most likely the drawn samples are from Truncated Distribution.

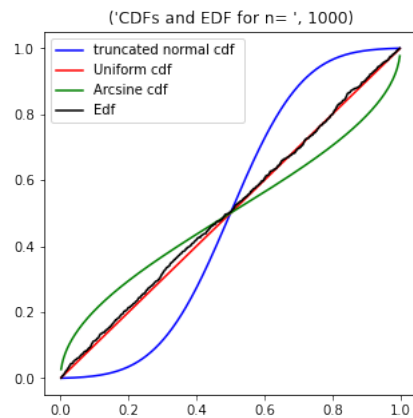
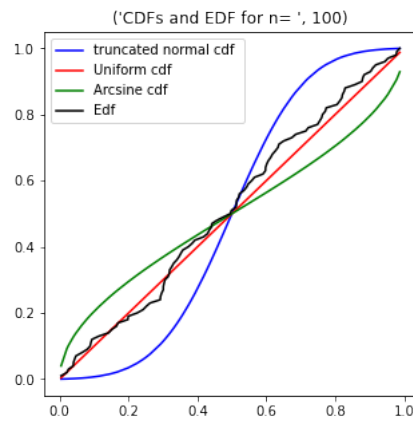


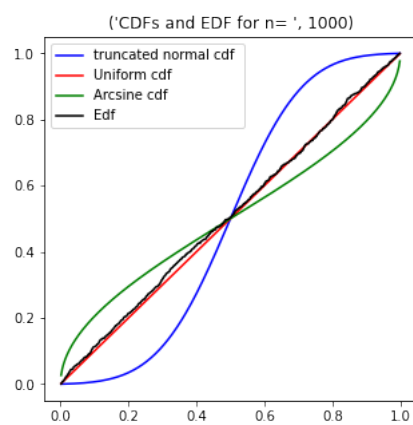
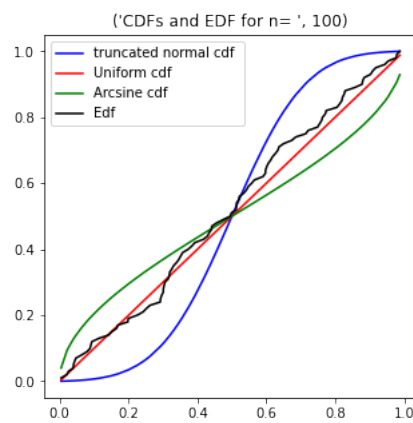
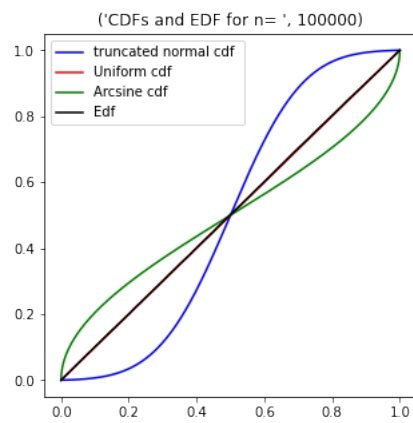
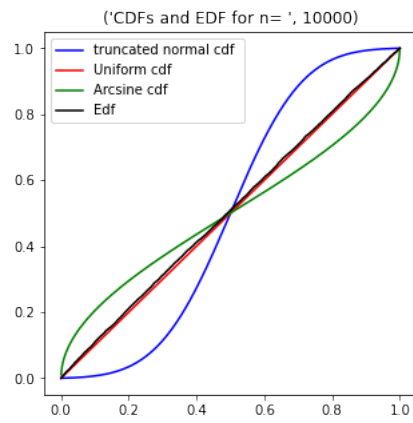


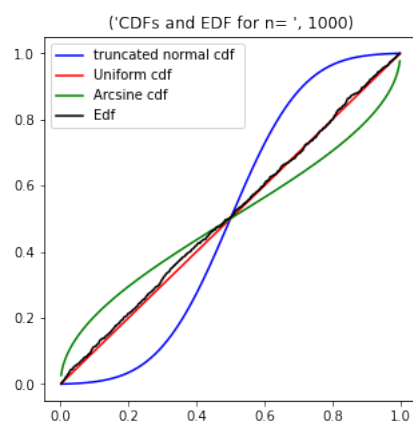
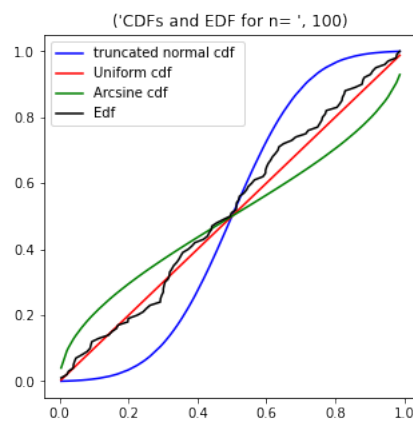
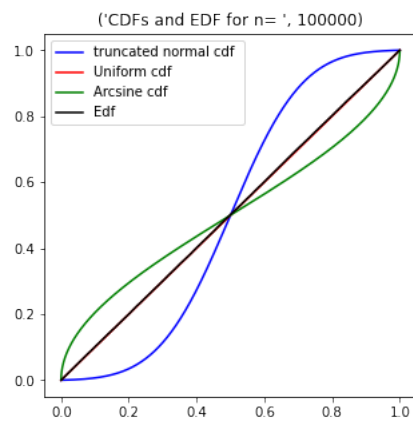
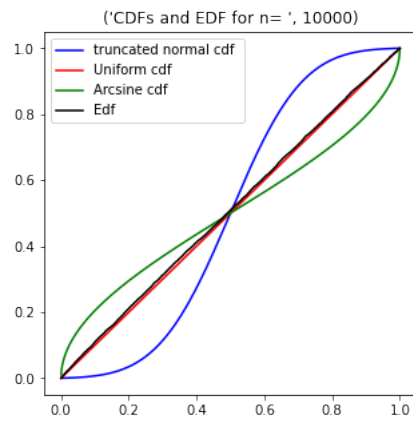


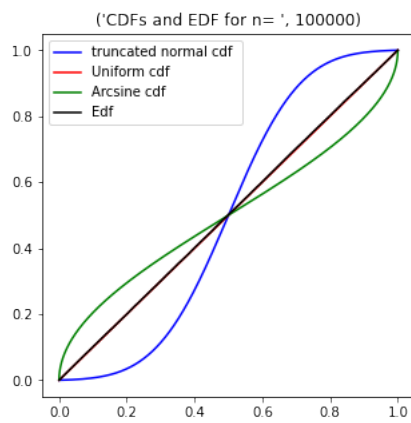
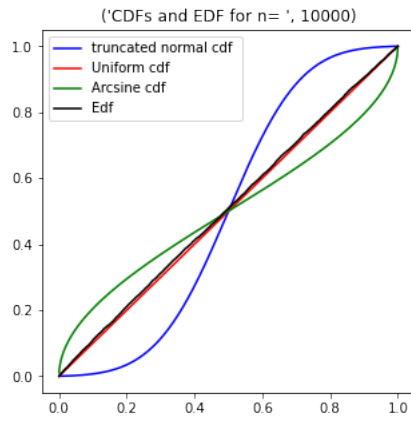


Samples X are drawn from Uniform Distribution $U(0, 1)$ and EDF is plotted along with CDFs corresponding to X assuming it to be drawn from Truncated Normal, Uniform and Arcsine distribution. Here we tested for 3 trials, and see that the EDF (empirical distribution function) is nearly close to Uniform CDF. Hence most likely the drawn samples are from Uniform Distribution.

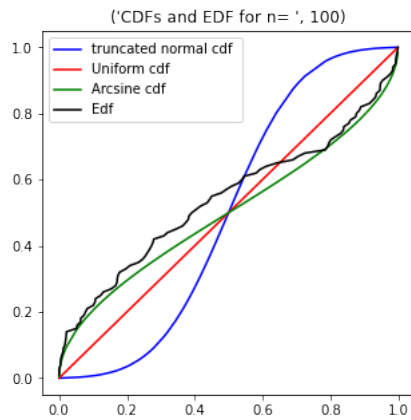


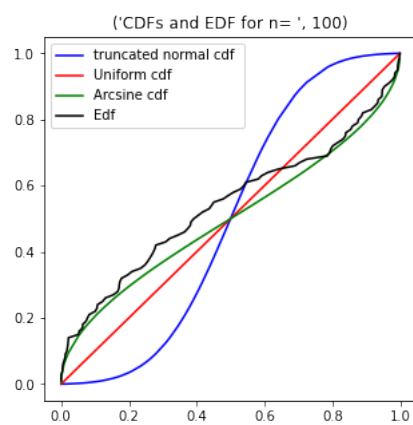
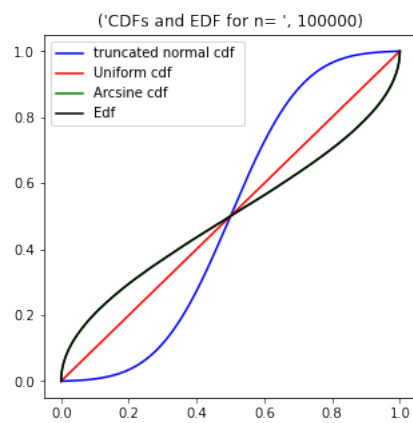
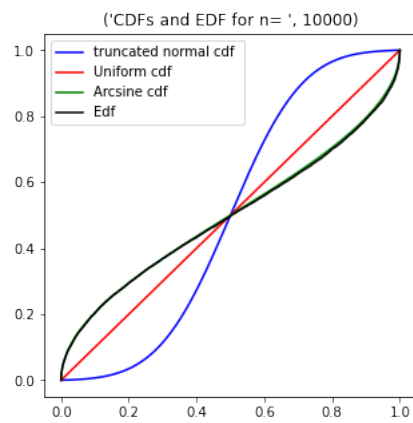
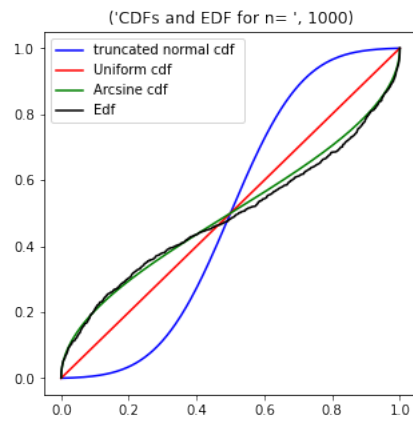


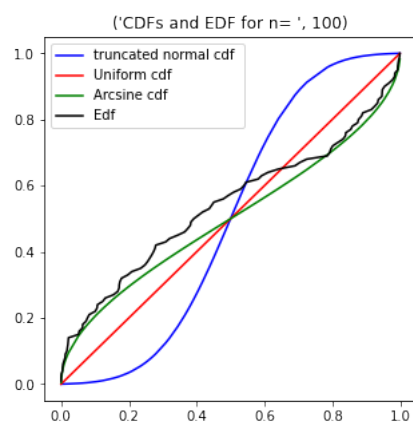
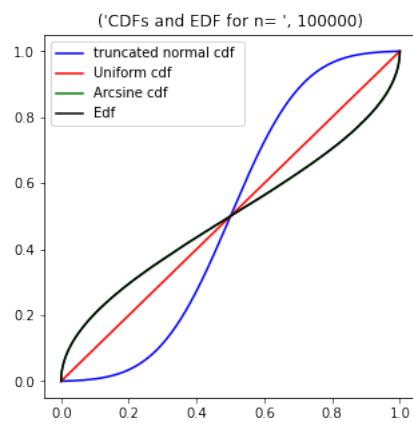
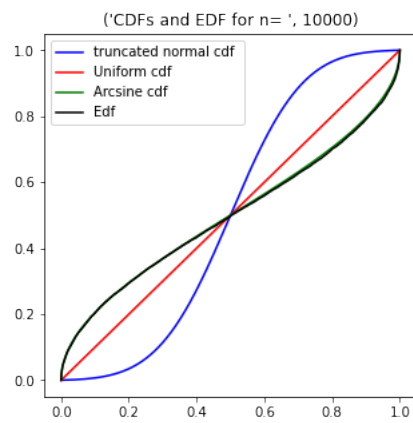
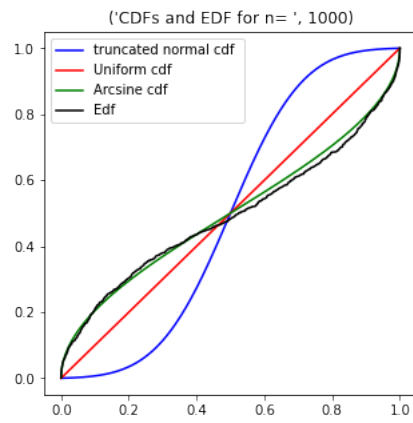


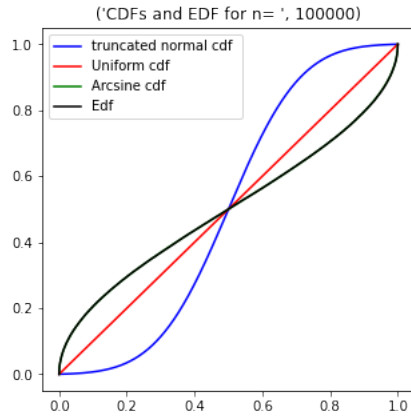
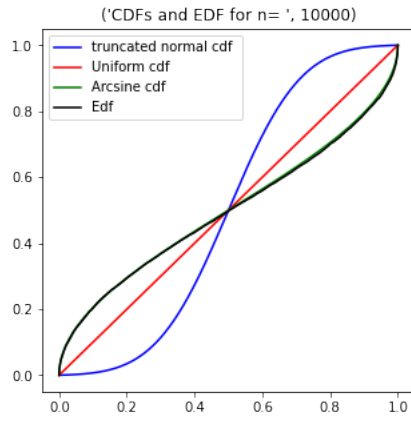
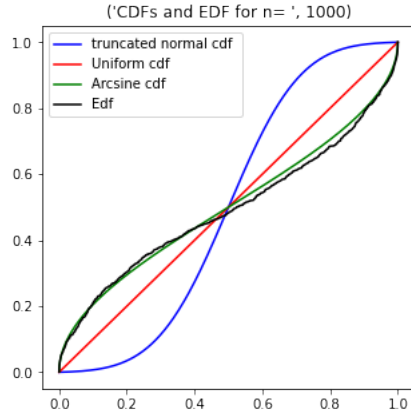


Samples X are drawn from Arcsine Distribution $Beta(1/2, 1/2)$ and EDF is plotted along with CDFs corresponding to X assuming it to be drawn from Truncated Normal, Uniform and Arcsine distribution. Here we tested for 3 trials, and see that the EDF (empirical distribution function) is nearly close to Arcsine CDF. Hence most likely the drawn samples are from Arcsine Distribution.









Here we use the power model relation between K-S statistics D and Sample size (n) given by

$$\epsilon(n) = C \times n^p$$

Where C and p are proportionality coefficient and power index respectively.

Taking log base 10 on both sides we get $\log_{10}(D) = \log_{10}(C) + p \times \log_{10}(n)$

Evaluating C and p based on different distribution using curve fitting,

For truncated normal distribution we got $\epsilon(n) = 0.07897475774365889 \times n^{-0.3066345554804011}$

For Uniform distribution we get $\epsilon(n) = 0.33203606128784174 \times n^{(-0.13316969540722856)}$.

For arcsine distribution we get $\epsilon(n) = 0.35788293117247916 \times n^{(-0.14863630206401224)}$

Plotting Average K-S statistics vs Sample size (n) on \log_{10} scale for all 3 distributions, we can see that KS statistic decreases with the increasing value of sample size n .

