

1.

2013/MANG SCI. PAPER 1 Q3

DAVID WEITBRECHT.

Random number generators (RNG's) ^{ideally} yield values y that are uniformly distributed on $(0,1)$ and are independent. They are "input" to system that transform them.

- A. Use sketches and formulae for the pdf and the cdf to explain what is meant by saying y is uniformly distributed on $(0,1)$.

pdf

If the distribution is described as uniform on $(0,1)$, the probability density function associated with this experiment should be constant on $(0,1)$.

A continuous random variable has a uniform distribution on the interval (a,b) if its probability density function $f(x)$ is given by $f(x)=0$, if x is not in $[a,b]$ and $f(x) = \frac{1}{b-a}$ for $a \leq x \leq b$.

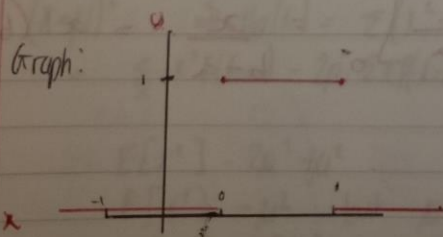
We denote this distribution $U(a,b)$

general cdf
$$\begin{cases} 0 & x \leq 0 \\ x & 0 \leq x \leq 1 \\ 1 & x \geq 1 \end{cases}$$

In our case: $a=0, b=1$

$f(x) = \frac{1}{1-0} = 1 = 1$

Our cdf
$$\begin{cases} 0 & x \leq 0 \\ 1 & 0 \leq x \leq 1 \\ 1 & x \geq 1 \end{cases}$$



We need pdf to find probability of a continuous random variable lying with some interval (a,b) . This lead to cdf.

3

2013 MAUG SCI PAPER 1 Q3 DAVID WETTBRECH

- (3a) Show how calculus can be used to compute $E[Y]$, $E[Y^2]$ and $\text{var}[Y]$ explaining the meanings of each term.

 $E[Y]$ Expected value of Y , the long run average value

$$E(Y) = \int_{-\infty}^{\infty} y f(y) dy$$

$$= \int_0^b y \left(\frac{1}{b-a}\right) dy$$

$$= \frac{1}{b-a} \int_0^b y dy$$

$$= \frac{1}{b-a} \left(\frac{1}{2}\right)(y^2) \Big|_{y=0}^b$$

$$= \frac{1}{b-a} \left(\frac{1}{2}\right)(b^2 - 0^2) = \frac{1}{2} \frac{(b-a)(b+a)}{(b-a)} = \frac{b+a}{2} \Rightarrow \frac{1}{2} \quad \checkmark$$

$$E(Y^2) = \frac{1}{b-a} \int_0^b y^2 dy$$

long run average of Y^2

$$= \frac{1}{b-a} \left[\frac{y^3}{3}\right]_{y=0}^b$$

$$= \frac{1}{b-a} \left[\frac{b^3 - 0^3}{3}\right] = \frac{1}{3} \frac{(b^3 - 0^3)}{(b-a)} = \frac{1}{3} \frac{(b^2 + b + 0)}{(1-0)} = \frac{1}{3} \quad \checkmark$$

VarianceVariance is a measure of the degree of dispersion of X about its expected value $E(X)$ ($E(X)$ is mean of X)

$$(X - \mu)^2 \Rightarrow \text{Var}(X) = \sum (x^2 - 2\mu x + \mu^2) P(X=x)$$

$$= \sum x^2 P(X=x) - 2\mu \sum x P(X=x) + \mu^2 \sum P(X=x)$$

$$* E(X) = \sum x P(X=x) \text{ and } \sum P(X=x) = 1$$

$$E[X^2] - 2\mu^2 + \mu^2$$

$$E[X^2] - \mu^2 \text{ but } \mu = E(X)$$

$$E[X^2] - E[X]^2$$

$$\Rightarrow \frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12} \quad \checkmark$$

4
3(Aii) Discuss independence in the context of the Pseudo-RNG's commonly used in Software

A truly random number is independent of any other random numbers, or anything at all.

Pseudo random numbers are deterministic, they are NOT independent.

Pseudo random numbers are produced using random number before, and inputting it into a complex algorithm.

Easily reproduced when we know algorithm.

For our purposes, Pseudo RNG are "statistically" random for any of our experiment.

3(B) The Transformation $X = -\ln(1-y)$ yields valid for a variable X that is said to have an Exponential Distribution with expected value of 1. Use an event identity to determine algebraically the cdf for X and hence its pdf. Explain with sketches how these can each be used to compute $\Pr(X \leq 1)$.

We work our function in terms of X .

$$X = -\ln(1-y)$$

$$-X = \ln(1-y)$$

$$e^{-X} = 1-y$$

$$y = 1 - e^{-X}$$

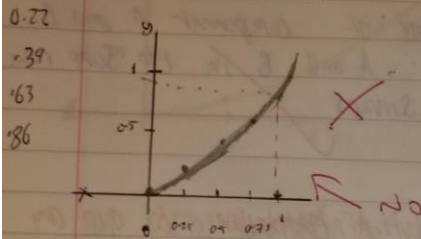
this is our cdf.

$$\text{cdf} = \begin{cases} 0 & X \leq 0 \\ 1 - e^{-X} & 0 \leq X < \infty \\ 1 & X \geq 1 \end{cases}$$

$$\text{e.g. } F_X(2) = 1 - e^{-2} \neq 1$$

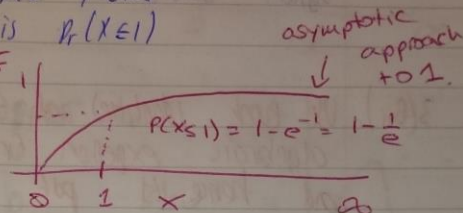
5. 2013 MANG SCI PAPER 1 Q3 DAVID WEITBRECHT

Graph of cdf.



$P_r(X \leq 1)$ from cdf.

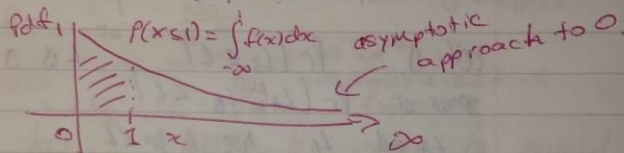
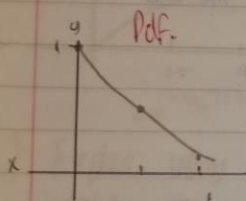
Use $x=1$, bring the vertical line to hit the curve, bring the horizontal line to meet the y-axis. This is $P_r(X \leq 1)$.



Pdf is cdf differential

$$\frac{d(1-e^{-x})}{dx} = f(x) = e^{-x}$$

$$\begin{cases} 0 \leq x \\ 0 < \infty \end{cases}$$



1. A system has 2 components that run in parallel. Providing that either are working, the system works. The lifetimes are exponentially distributed with expected value 1 and are taken as independent. Explain using spreadsheet ideas of rows and columns how to simulate system life time.

	lifetime A	lifetime B	Max of A or B
1	$-\ln(1-\text{RAND}())$	$-\ln(1-\text{RAND}())$	$= \text{Max}('A1:B1')$
2	$-\ln(1-\text{RAND}())$	$-\ln(1-\text{RAND}())$	$= \text{Max}('A2:B2')$
⋮	⋮	⋮	⋮
n	$-\ln(1-\text{RAND}())$	$-\ln(1-\text{RAND}())$	$= \text{MAX}('A n: B n')$

Pick n similarly large and simulate as above.

To find average system lifetime, add all values in

in "C" column and divide by count.

- Column A and B represent the lifetime of component A and B.
- Column C is max of both A and B in the same row.
- System survives if A or B survives.

3(Bii) Use Euler Identity and Exponential distribution to give an algebraic expression for the cdf of the system lifetime and hence its pdf. Sketch both functions.

L_s = lifetime system L_A = lifetime A L_B = lifetime B

cdf is $\Pr(L_s \leq L)$ L is a constant.

max of $\Pr(L_A \text{ AND } L_B \leq L)$

L_A and L_B have to be $\leq L$

$\Pr(L_A \leq L)$ $\Pr(L_B \leq L)$

lifetime of A and B are $-\ln(1-y)$ No. $\Pr(L_s \leq L) = 1 - e^{-L}$

$\Rightarrow \Pr(-\ln(1-y_A) \leq L)$ and $\Pr(-\ln(1-y_B) \leq L)$ $\Pr(L_s \leq L)$

\Rightarrow manipulate like earlier No.

$\Pr(y_A \leq 1 - e^{-L})$ and $\Pr(y_B \leq 1 - e^{-L})$

$\Pr(y_A, y_B) = (1 - e^{-L})^2 (1 - e^{-L})$

$1 - e^{-L} - e^{-L} + e^{-2L}$

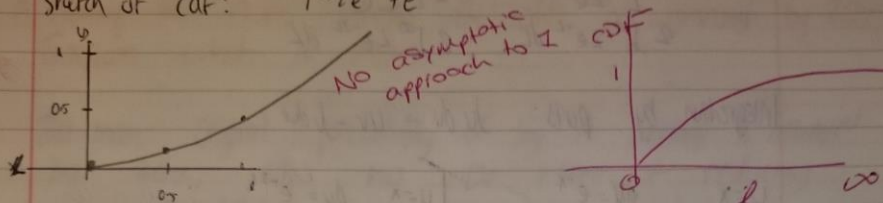
$\Pr(y_A, y_B) \leq (1 - 2e^{-L} + e^{-2L})$

y_A and y_B are random numbers between 0 and 1.

cdf $\Rightarrow \begin{cases} 0 & L \leq 0 \\ 1 - 2e^{-L} + e^{-2L} & 0 \leq L \leq 1 \\ 1 & L \geq 1 \end{cases}$

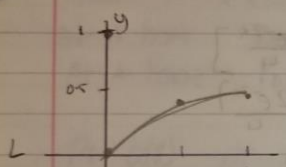
7. 2013 MANG XI PAPER 1 Q3 DAVID WEITBRECHT.

Sketch of cdf: $1 - 2e^{-L} + e^{-2L}$

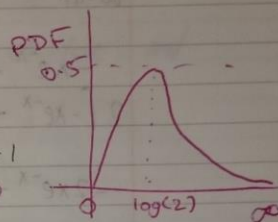


No asymptotic approach to 1 cdf

$$pdf = f(x) = \frac{d}{dL}(1 - 2e^{-L} + e^{-2L}) = 2e^{-L} - 2e^{-2L}$$



$$pdf = \begin{cases} 0 & L < 0 \\ 2e^{-L} - 2e^{-2L} & 0 \leq L \leq 1 \\ 0 & L > 1 \end{cases}$$



$$\begin{aligned} \frac{d}{dx}(2e^{-L} - 2e^{-2L}) &= -2e^{-L} + 4e^{-2L} = 0 \\ \Rightarrow 4e^{-2L} &= 2e^{-L} \Rightarrow \log(4) - 2L = -L \\ \log(4) &= L \\ 2e^{-\log(2)} - 2e^{-2\log(2)} &= 2 \times \frac{1}{2} - 2 \times \frac{1}{4} = 1 - \frac{1}{2} = \frac{1}{2} \end{aligned}$$

3(biii) Explain using calculus how in principle to compute the expected value of system lifetime. Will this be greater, or less than 1, the expected lifetime of a computer? Why?

To calculate expected value we would integrate pdf.

$$\begin{aligned} E[L] &= \int_0^{\infty} L(2e^{-L} - 2e^{-2L}) dL \\ &= \int_0^{\infty} 2Le^{-L} dL - \int_0^{\infty} 2Le^{-2L} dL \end{aligned}$$

The expected value of variable L with prior $\lambda = 1$ is

$$E[L] = \frac{1}{\lambda} = 1$$

NOT SURE HOW TO DO THAT

$$2 \int_0^{\infty} L e^{-L} dL - 2 \int_0^{\infty} L e^{-2L} dL$$

$$\begin{aligned} 2e^{-L} &= -e^{-L} \\ u = 2e^{-L} & \quad e^{-L} dL = du \\ du &= -e^{-L} dL \quad v = -e^{-L} \\ -e^{-L} &= u \quad + e^{-L} du \\ 2(-e^{-L}) &+ 2e^{-L} - 2e^{-L} + 2e^{-L} \end{aligned}$$

$$\mathcal{L} 2e^{-L} - \mathcal{L} 2e^{-2L}$$

$$2 \int_0^{\infty} L e^{-L} dL - 2 \int_0^{\infty} L e^{-2L} dL$$

Integration by parts: $\int u dv = uv - \int v du$

$$u = x \quad dy = e^{-x}$$

$$du = dx \quad y = -e^{-x}$$

$$u = x \quad dv = e^{-2x}$$

$$du = dx \quad v = \int e^{-2x} = -\frac{e^{-2x}}{2}$$

$$-x e^{-x} - \int -e^{-x} dx \quad \left| \quad 2 \left[-\frac{x e^{-2x}}{2} - \int -\frac{e^{-2x}}{2} dx \right] \right.$$

$$2 \left[-x e^{-x} - e^{-x} \right] \quad \left| \quad 2 \left[-\frac{x e^{-2x}}{2} - \frac{e^{-2x}}{4} \right] \right.$$

$$-2x e^{-x} - 2e^{-x} \quad - \quad \left(-\frac{2x e^{-2x}}{2} \right) - \left(-\frac{2e^{-2x}}{4} \right)$$

$$-2x e^{-x} - 2e^{-x} + x e^{-2x} + \frac{2e^{-2x}}{2} \Bigg|_{x=0}^{\infty}$$

$$-2(0)e^0 - 2e^0 + (0)e^0 + \frac{2e^0}{2} \quad \text{at } x=0.$$

$$-2 + 1 = -\frac{3}{2}$$

$$\text{at } x=\infty \quad e^{-x} \Rightarrow 0 \Rightarrow = 0$$

$$0 - \left(-\frac{3}{2}\right) = \frac{3}{2}$$

$$L \text{ is } > 1 \quad \text{it is } \frac{3}{2}$$



Q4 A. Playing at home 'increases' the odds of winning by 20%. Explain what it means.

This means original probability of say A beating B increased by 20%, this has the effect of reducing B's probability of winning away by 20%.

For example if only two outcomes, win or lose and $P_{A \text{ win}} = P_{B \text{ win}} = 0.5$.

At A home $P_{A \text{ win}} = 0.5 \times 1.2 = 0.6$

At A home $P_{B \text{ win}} = 0.5 \times 0.8 = 0.4$

No.

The odds are now in favor of A at home in the ratio of 3:2

A Home.

$$\frac{P(A \text{ win})}{P(B \text{ win})} = \frac{1 \times 1.2}{1 \times 0.8} = \frac{1.2}{0.8} = \frac{3}{2}$$

$$P = 1.2(1-p) = 1.2 - 1.2p \quad 2.2p = 1.2 \quad p = \frac{1.2}{2.2} = 0.55$$

4 b. It is known that A beats B. Explain how to compute the probability that this was a home game. Explain both the probability and odds form of Bayes' Rule.

Here we want $P(A_{\text{home}} | A_{\text{win}})$

Using Bayes' rule we get $P(A_{\text{win}} | A_{\text{home}}) = \frac{P(A_{\text{home}} \text{ and } A_{\text{win}})}{P(A_{\text{home}})}$

This is a conditional probability which reflects our knowledge of the occurrence of A at home.

The usual probability form of Bayes' rule is $P(A|B) = \frac{P(A \cap B)}{P(B)}$ if $P(B) > 0$.

$$P(A_{\text{home}} | A_{\text{win}}) = \frac{P(A_{\text{win}} | A_{\text{home}}) P(A_{\text{home}})}{P(A_{\text{win}})} = \frac{0.55 \times 0.5}{(0.55 \times 0.5) + (0.45 \times 0.5)} = 0.55$$

2

Bayes' rule specifies how probabilities must be updated in the light of new information.

Before knowing an event happened we assign prior probabilities $P(H)$ and $P(\bar{H}) = 1 - P(H)$, where H and \bar{H} are mutually exclusive.

How do the prior probabilities change once evidence in the form of the knowledge that the event E has occurred becomes available?

The updated value of the probability that the hypothesis H is true given the fact that E occurred is denoted $P(H|E)$.

To calculate posterior probability $P(H|E)$ we use Bayes' rule.

We use odds.

$$\frac{P(A \text{ Home} | A \text{ win})}{P(A \text{ Away} | A \text{ win})} = \frac{P(A \text{ Home})}{P(A \text{ Away})} \cdot \frac{P(A \text{ win} | A \text{ Home})}{P(A \text{ win} | A \text{ Away})}$$

$$= \frac{P(A \text{ win} | A \text{ Home})}{P(A \text{ win} | A \text{ Away})} \cdot \frac{P(A \text{ Home})}{P(A \text{ Away})} = \frac{0.55}{0.45} \cdot \frac{1.222}{1.222} = 1.222$$

Posterior probability $P(H|E)$ satisfies: $\frac{P(H|E)}{P(\bar{H}|E)} = \frac{P(H)}{P(\bar{H})} \cdot \frac{P(E|H)}{P(E|\bar{H})}$

In words, Bayes' rule in odds form states: $\frac{P}{1-P} = \frac{P}{1-P} \cdot \frac{P}{1-P}$

Posterior odds = prior odds \times likelihood ratio $P = 0.2225 - 0.2225$

Factor $\frac{P(H)}{P(\bar{H})}$ gives prior odds in favor of H occurring $P = \frac{0.2225}{1.222} = 0.55$

The second fraction represents the effect E will have on the belief H happened.

- C i. In these circumstances the number of games N won by any team follows, marginally, a binomial distribution. Explain what this means and the key assumption that it is based on. What is the variance of the number of games won by each team?

3

2013 MANG XI PAPER 1 Q4 DAVID WEITBRECHT.

4c i. Binomial $Y \sim B(n, p)$

Basic Assumptions:

1. Y are positive integers: $0, 1, 2, \dots, n$
2. Finite number of trials n
3. Only two outcomes "success" and "failure."
4. Independent trials. \rightarrow with same probability of success in each trial.

Our situation satisfies these conditions, Y is 0, 1 or 2, 3 or 4. $N = 4$ = number of games

Each game independent of each other.

P success for each trial is the same

$$P_r(Y=y) = \binom{n}{y} p^y (1-p)^{n-y}$$

$$P_r(Y=0) = \binom{4}{0} (0.5)^0 (0.5)^4 = 1/16$$

$$P_r(Y=1) = \binom{4}{1} (0.5)^1 (0.5)^3 = 1/4$$

$$P_r(Y=2) = \binom{4}{2} (0.5)^2 (0.5)^2 = 3/8$$

$$P_r(Y=3) = \binom{4}{3} (0.5)^3 (0.5)^1 = 1/4$$

$$P_r(Y=4) = \binom{4}{4} (0.5)^4 (0.5)^0 = 1/16$$

$$= 1$$

Variance calculated by $np(1-p)$ for binomial dist

$$4(0.5)(1-0.5) = 1.$$



- 4cii. Suppose probability of a draw in every game is 0.1. Wins attract 3 points, draws 1 point, and losses no points. Explain why binomial distribution no longer applied to the distribution of the number of points won by each of the teams. Show it is simple to compute the variance of this random variable.

4

4c (ii) Basic assumption has been broken with introduction of third probability, we have 3 outcomes not 2.
 \Rightarrow Binomial no longer applies

No longer
Bernoulli trials

	pr	value (x)	x^2
win	0.45	3	9
lose	0.45	0	0
draw	0.1	1	1

$$E[x] = 0.45 \times 3 + 0.45 \times 0 + 0.1 \times 1 = 1.45 \text{ points}$$

$$E[x^2] = 0.45 \times 9 + 0.45 \times 0 + 0.1 \times 1 = 4.15 \text{ points}$$

$$\text{Variance} = E[x^2] - E[x]^2$$

$$4.15 - (1.45)^2 = 2.0475 \text{ points this is for 1 game.}$$

For 4 games it is 4×2.0475

iii. If home advantage is 20% as above, and draws are as described, what is the variance of the number of points won by each of the team

	home	away
Pr win	0.54	0.36
lose	0.36	0.54
draw	0.1	0.1

SEE

SHEET

6



Var(home)

$$E[x^2] - E[x]^2$$

$$[0.54 \times 9 + 0.1 \times 1] - [0.54 \times 3 + 0.1]^2$$

$$= 2.0016$$

+

$$= 3.9492$$

away

$$E(x) = 1.18 \quad E(x^2) = 3.84$$

$$\text{Var} = 1.9476$$

5. 2013 MANG SCI PAPER 1 Q4 DAVID WEITBRECHT

d. Understanding uncertainty site showed that for the premier league with 20 teams and 38 matches, theoretical variance of number of points gained is 61; computed in a way similar to (iii), in 2007 computed variance across 20 teams was 239. The site concludes "about half the spread of points is due to chance alone".

Assuming all teams were of equal standard, so that each match resulted in a home win with probability p_H , draw with p_D and away win with p_A where $p_H + p_A + p_D = 1$

X_H is number of points a team wins in a home match, variance of X_H is

$$E(X_H) = 3p_H + p_D$$

$$E(X_H^2) = 9p_H + p_D$$

$$\begin{aligned} \text{Var} &= E(X_H^2) - E(X_H)^2 \\ &= 9p_H + p_D - (9p_H^2 + p_D^2 + 6p_H p_D) \\ &= 9p_H + p_D - E(X_H)^2 \end{aligned}$$

Same for A.

$N/2$ home games, $N/2$ away games

Total points T is sum of all points in the individual matches

$$\begin{aligned} \text{Expected points } m_T &= E(T) = \frac{N}{2}(m_H + m_A) = \frac{N}{2}(3(p_H + p_A) + 2p_D) \\ &= \frac{N}{2}(3 - p_D) \end{aligned}$$

$$\begin{aligned} \text{Variance } V(T) &= \frac{N}{2}(V_H + V_A) = \frac{N}{2}(9(p_H + p_A) + 2p_D - m_H^2 - m_A^2) \\ &= \frac{N}{2}(9 - 7p_D - m_H^2 - m_A^2) \end{aligned}$$

6

This is how variance = 61.

The variance of the actual league points at the end of season was 234 compared with theoretical variance of 61 were all teams all of equal quality and the results of matches were due to chance. *half in 2!*

Since $61/234 = 26$, we conclude that 26% of the variance in the premier league points is due to chance.

$$SD = \sqrt{61} \approx 7.8 \leftarrow \text{spread of points according to model}$$

$$SD = \sqrt{234} \approx 15.5 \leftarrow \text{total spread of points (only 7.8 explained by model)}$$

4 (iii)

	home	away
win	0.54 $\nearrow 0.9 \times 0.55$	0.36 0.9×0.45
lose	0.36 0.9×0.45	0.54 0.9×0.55
draw	0.1 $\searrow 0.1$	0.1 0.1

$$E(x) = 0.54 \times 3 + 0.1$$

$$1.72$$

$$E(x^2) = 0.54 \times 9 + 0.1$$

$$4.96$$

$$Var = 4.96 - (1.72)^2$$

$$2.0016$$

$$E(x) = 0.36 \times 3 + 0.1$$

$$1.18$$

$$E(x^2) = 0.36 \times 9 + 0.1$$

$$3.34$$

$$Var = 3.34 - (1.18)^2$$

$$1.9476$$

otherwise fine.

Variance total points $n=4$ is $\frac{n}{2}(V_A + V_H)$

$$\frac{4}{2}(2.0016 + 1.9476)$$

$$= 7.8984 \text{ points}$$

2013 MANGSCI PAPER 1 Q3 CORRECTION

3ai. $E[y]$ Expected value of y , the long run average value

$$E[y] = \int_{-\infty}^{\infty} y f(y) dy$$

$$= \int_0^b y \left(\frac{1}{b-a}\right) dy$$

$$\frac{1}{b-a} \frac{y^2}{2} \Big|_0^b = \frac{1}{2} \frac{1}{b-a} (b^2 - a^2) = \frac{1}{2} \frac{(b-a)(b+a)}{b-a} = \frac{b+a}{2} = \frac{1}{2}$$

 $E[y^2]$ long run average of y^2

$$\frac{1}{b-a} \int_0^b y^2 dy$$

$$\frac{1}{b-a} \left[\frac{y^3}{3} \right] \Big|_0^b = \frac{1}{b-a} \left[\frac{b^3 - a^3}{3} \right] = \frac{1}{3} \left[\frac{b^3 - a^3}{b-a} \right] = \frac{1}{3} \left[\frac{b^2 + ba + a^2}{1} \right] = \frac{1}{3}$$

Variance

Variance is a measure of the degree of dispersion of X about its expectation $E[X]$ $E[X]$ is mean of X

$$(X - \mu)^2 \Rightarrow \text{Var}(X) = \sum (X^2 - 2\mu X + \mu^2) p(X=x)$$

$$= \sum x^2 p(X=x) - 2\mu \sum x p(X=x) + \mu^2 \sum p(X=x)$$

$$E[X^2] - 2\mu E[X] + \mu^2 (1) \quad \mu = E[X]$$

$$E[X^2] - E[X]^2$$

$$\frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}$$

A ii A truly random number is important of any other random number or anything of that.

Pseudo random numbers are deterministic, they are not independent

Pseudo random numbers are produced using random numbers before and inputting it into a complex algorithm

2

They are easily reproduced when we know the algorithm

For our purpose, pseudo RNGs are "statistically" random for any of our experiments

b.i. we want our function in terms of $y = "$

$$x = -\ln(1-y)$$

$$-x = \ln(1-y)$$

$$e^{-x} = 1-y$$

$$e^{-x} - 1 = -y$$

$$y = 1 - e^{-x} \quad \text{this is our cdf}$$

$$\begin{cases} 0 & 0 \leq x \\ 1 - e^{-x} & 0 < x \leq \infty \end{cases}$$

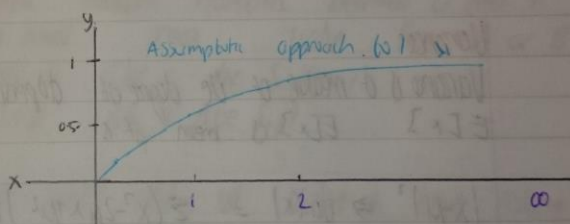
Graph of cdf.

$$x = 0.25 \quad y = 0.22$$

$$0.5 \quad 0.39$$

$$1 \quad 0.63$$

$$2 \quad 0.86$$



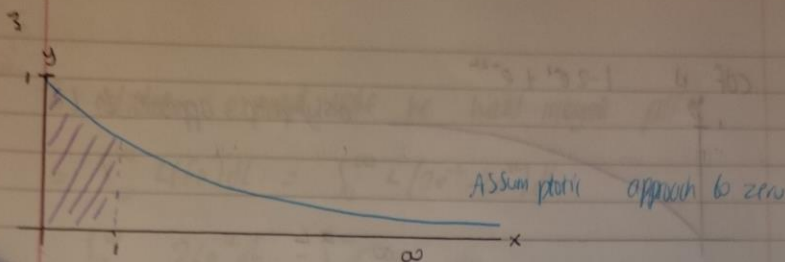
$$\Pr(X \leq 1) \text{ from cdf} = 1 - e^{-1} = 1 - \frac{1}{e} = 0.63$$

or bring the vertical until it hits the curve and bring it horizontal and read off value

$$\text{Pdf is cdf derivative}$$

$$\frac{d(1 - e^{-x})}{dx} = e^{-x}$$

$$\begin{cases} e^{-x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$



$Pr(X \leq 1)$ is area to left of one or $e^{-1} = \frac{1}{e} = 0.368$

Life time A	Life time B	Max of A and B
$-\ln(1-RAND)$	$-\ln(1-RAND)$	$= \text{Max}(A, B)$
\vdots	\vdots	\vdots
$-\ln(1-RAND)$	$-\ln(1-RAND)$	$= \text{Max}(A_N, B_N)$

- Use Rand() to generate random numbers between 0 and 1.
- In column C put record max value from A or B as if one component of the system is
- (Choose N suitably large (1000 or more say))

- To find average system lifetime, add all values in third column and divide by count.
- System survives if A or B survives

bii. L_S - lifetime system L_A - lifetime A L_B - lifetime B

cdf is $Pr(L_S \leq L)$ L is constant
max of $Pr(L_A \text{ and } L_B \leq L)$

L_A and L_B must be $\leq L$

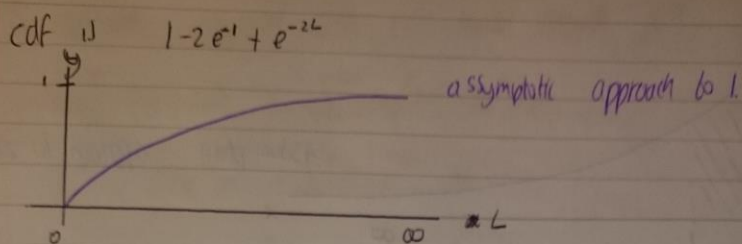
$Pr(L_A \leq L)$ and $Pr(L_B \leq L)$

$$= (1 - e^{-L})(1 - e^{-L})$$

$$1 - e^{-L} - e^{-L} + e^{-2L} = 1 - 2e^{-L} + e^{-2L} \text{ for } L \geq 0$$

0 otherwise

4



pdf is $\frac{d(1 - 2e^{-L} + e^{-2L})}{dL} = 2e^{-L} - 2e^{-2L}$

at max/min $\frac{d}{dL} = 0 \quad -2e^{-L} + 4e^{-2L} = 0$

$$4e^{-2L} = 2e^{-L}$$

$$2e^{-2L} = e^{-L}$$

$$2 \ln(2e^{-2L}) = \ln(2e^{-L})$$

$$\ln(2) - 2L = -L$$

$$\ln(2) = L \text{ at max/min}$$

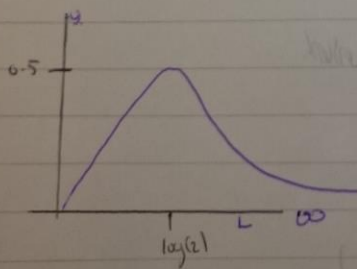
into pdf:

$$2e^{-\ln(2)} - 2e^{-2\ln(2)}$$

$$2\left(\frac{1}{2}\right) - 2\left(\frac{1}{4}\right) = 1 - \frac{1}{2} = \frac{1}{2}$$

at $\log(2)$ max value of $\frac{1}{2}$

pdf = $\begin{cases} 0 & \text{otherwise} \\ 2e^{-L} - 2e^{-2L} & L \geq 0 \end{cases}$



5

biii.

To calculate expected value we will integrate pdf.

$$= \int_0^{\infty} L(f(x)) dx = \int_0^{\infty} L(2e^{-x} - 2e^{-2x}) dx$$

$$\int_0^{\infty} 2Le^{-x} dx - \int_0^{\infty} 2Le^{-2x} dx$$

Integration by parts $\int u dv = u \cdot v - \int v du$

$$u = x \quad du = dx$$

$$dv = e^{-x} \quad v = -e^{-x}$$

$$u = x \quad du = dx$$

$$dv = e^{-2x} \quad v = -\frac{e^{-2x}}{2}$$

$$= -Le^{-x} - \int -e^{-x} dx$$

$$= -Le^{-x} - e^{-x}$$

$$-\frac{Le^{-2x}}{2} - \int \frac{-e^{-2x}}{2} dx$$

$$\left[-2Le^{-x} - 2e^{-x} + \frac{2Le^{-2x}}{2} + \frac{2e^{-2x}}{4} \right]_0^{\infty}$$

$$-2(\infty)e^{-\infty} - 2e^{-\infty} + \frac{2Le^{-\infty}}{2} + \frac{2e^{-\infty}}{4} - [0 - 2(1) + 0 + \frac{2}{4}]$$

$$= \frac{3}{2} = 1.5$$

$$L(1) > 1 \quad \text{if } (1) \quad 3/2$$

1 2013 MANG XI PAPER Q4 CORRECTED

Q4A. $P(A \text{ win}) = 1 \times 1.2 = 1.2$

$$P(A \text{ win}) = 1.2 P(B \text{ win})$$

$$p = 1.2(1-p) = (1.2 - 1.2p)$$

$$1.2(1-p) \quad 1.2 - 2.2p = 0$$

$$2.2p = 1.2$$

$$2.2p = 1.2$$

$$p = \frac{1.2}{2.2} = 0.55$$

$$p = 0.55$$

Probably A win = 0.55

Probably B win = 0.45

B. Here we want $P(A \text{ lose} | A \text{ wins})$

Using Bayes rule we get

$$P(A \text{ lose} | A \text{ win}) = \frac{P(A \text{ win} | A \text{ lose}) P(A \text{ lose})}{P(A \text{ win})}$$

$$= \frac{0.55 \times 0.5}{0.55 \times 0.5 + 0.45 \times 0.5}$$

$$= 0.55 \text{ chance}$$

By manipulation $P(A|B) = \frac{P(AB)}{P(B)}$

$$P(AB) = P(A|B)P(B)$$

Bayes rule specifies how probabilities must be updated in light of new information

Before knowing an event happened we assign prior probabilities $P(H)$ and $P(\bar{H}) = 1 - P(H)$ where H and \bar{H} are mutually exclusive

The updated value of the probability that of

2

hypothesis is true given the fact that E occurs is denoted $P(H|E)$.

To calculate posterior probability $P(A_{\text{home}} | A_{\text{win}})$ we use Bayes' rule

$$\frac{P(A_{\text{home}} | A_{\text{win}})}{P(A_{\text{away}} | A_{\text{win}})} = \frac{P(A_{\text{home}})}{P(A_{\text{away}})} \cdot \frac{P(A_{\text{win}} | A_{\text{home}})}{P(A_{\text{win}} | A_{\text{away}})}$$

$$= \frac{0.55}{0.45} = 1.222$$

$$\frac{p}{1-p} = 1.222$$

$$p = 1.222 - 1.222p$$

$$2.222p = 1.222$$

$$p = \frac{1.222}{2.222} = 0.55$$

4. Binomial $Y \sim B(n, p)$

Basic assumptions

- Y are positive integers $0, 1, \dots, n$
- finite number of trials n
- Only two outcomes "Success" or failure
- Independent trials with same probability of success in each trial
- Our situation satisfies these conditions, $Y = 0, 1, 2, 3$ or 4
- $n = 4$ = number of games
- Each game independent of each other
- p Success for each trial is the same

$$P(Y=y) = \binom{n}{y} p^y (1-p)^{n-y}$$

$$P(Y=0) = \binom{4}{0} 0.5^0 (0.5)^4 = 1/16$$

$$P(Y=1) = \binom{4}{1} 0.5^1 (0.5)^3 = 1/4$$

$$P(Y=2) = \binom{4}{2} 0.5^2 (0.5)^2 = 3/8$$

$$P(Y=3) = \binom{4}{3} 0.5^3 (0.5)^1 = 1/4$$

$$P(Y=4) = \binom{4}{4} 0.5^4 (0.5)^0 = 1/16 = 1$$

Vercheck calculated by

$$np(1-p) = 4(0.5)(1-0.5) = 1$$

- 4ii. Basic assumptions have been breached with introduction of third probability, we have three outcomes not 2 as stated in our assumption
 \Rightarrow Binomial no longer applies as they are no longer Bernoulli trials

	pr	value x	x^2
win	0.45	3	9
lose	0.45	0	0
draw	0.1	1	1

$$E[x] = 0.45(3) + 0.45(0) + 0.1(1) = 1.45 \text{ points per game}$$

$$E[x^2] = 0.45(9) + 0.45(0) + 0.1(1) = 4.15 \text{ points}$$

$$\text{Variance} = E[x^2] - E[x]^2$$

$$= 4.15 - (1.45)^2 = 2.0075 \text{ points for 1 game}$$

$$4 \times 2.0075 \text{ is } 8.19 \text{ for 4 matches}$$

	here	away
win	0.4905	0.495
lose	0.405	0.495
draw	0.1	0.1

$$E[x] = 0.495(3) + 0.1(1) = 1.585$$

$$E[x^2] = 4.564$$

$$\text{var} = 2.051775$$

$$E[x] = 0.405(3) + 1(0.1) = 1.315$$

$$E[x^2] = 3.745$$

$$\text{var} = 2.015775$$

$$\frac{4}{2} (2.051775 + 2.015775)$$

$$= 8.135 \text{ points in 4 matches}$$

$$2 \text{ home } 2 \text{ away}$$

DAVID WEITBRECHT.

Q1 SECTION A

12 Variance of actual league points at the end of season was 239 compared with theoretical variance of 61, were all teams of equal quality and the result of match was due to chance

$SD = \sqrt{61} \approx 7.8 \leftarrow$ Spread of points according to model

$\sqrt{239} \approx 15.5 \leftarrow$ total spread of points

Only 7.8 of this is explained by model which is around half.