

MISC	PROB BASICS	CONT. DISTROS	COND. DISTROS VARS	EXPECTATIONS
Log $\ln(mn) = \ln(m) + \ln(n)$ $\ln\left(\frac{m}{n}\right) = \ln(m) - \ln(n)$ $\ln(m^r) = r \ln(m)$ Exponent $(ab)^x = a^x b^x$ $(a^x)^y = a^{xy}$ $a^x a^y = a^{x+y}$ Summation $\sum_{i=1}^n ar^{i-1} = a \frac{1-r^n}{1-r}$ Integrals $\int \frac{1}{x} dx = \ln x $ $\int e^{ax} dx = \frac{1}{a} e^{ax}$ $\int a^x dx = \frac{a^x}{\ln(a)}$ $\int \ln(x) dx = x \ln(x) - x$ $\int \cos(x) dx = \sin(x)$ $\int \sin(x) dx = -\cos(x)$ Derivatives $(e^x)' = e^x$ $(\ln(x))' = \frac{1}{x}$ $\sin(x) = \cos(x)$ $\cos(x) = -\sin(x)$ $(fg)' = fg' + f'g$ $\frac{1}{f} = -\frac{f'}{f^2}$ $(f(g(x)))' = f'(g(x))g'(x)$ $\frac{dy}{dx} = \frac{dy}{du} \frac{du}{dx}$	Properties $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ $P(A \cup B) \leq P(A) + P(B)$ Conditional $P(A \mid B) = \frac{P(A \cap B)}{P(B)}$ $P(A \cap B) = P(B)P(A \mid B)$ Total Prob Theorem $P(B) = P(A_1 \cap B) + \dots + P(A_n \cap B)$ $ = P(A_1)P(B \mid A_1) + \dots$ Bayes $P(A \mid B) = \frac{P(A)P(B \mid A)}{P(B)}$ Independence $P(A \mid B) = P(A)$ $P(A \cap B) = P(A)P(B)$	$P(a \leq x \leq b) = \int_a^b f_X(x) dx$ Disjoint $P(1 \leq x \leq 3 \text{ or } 4 \leq x \leq 5)$ $\phantom{P(1 \leq x \leq 3 \text{ or } 4 \leq x \leq 5)} = P(1 \leq x \leq 3) + P(4 \leq x \leq 5)$ Properties $f(x) \geq 0$ and $\int_{-\infty}^{\infty} f_X(x) dx = 1$ CDF Properties $\cdot \rightarrow_{x \rightarrow \infty} 1$ and $\rightarrow_{x \rightarrow -\infty} 0$ \cdot increasing/monotonic \cdot right-continuous $\text{---}^* \text{---}$ Uniform $f_X(x) = \frac{1}{b-a}$ $F_X(x) = \frac{x-a}{b-a}$ $E[X] = \frac{a+b}{2}$ $Var(X) = \frac{(b-a)^2}{12}$ Exponential <i>time to wait for something</i> $f_X(x) = \lambda e^{-\lambda x}$ $P(X \geq a) = \int_a^{\infty} \lambda e^{-\lambda x} dx = e^{-\lambda a}$ $F_X(x) = 1 - e^{-\lambda x}$ $E[X] = \frac{1}{\lambda}$ $E[X^2] = \frac{2}{\lambda^2}$ $Var(X) = \frac{1}{\lambda^2}$	MULTIPLE VARS $\sum_x \sum_y p_{X,Y}(x,y) = 1$ $P((X,Y) \in B) = \iint_{(x,y) \in B} f_{X,Y}(x,y) dx dy$ Marginals $p_X(x) = \sum_y p_{X,Y}(x,y)$ $f_X(x) = \int f_Y(y) f_{X Y}(x \mid y) dy$ Δ ranges: what values can Y take when X = x? $ = \int f_{X,Y}(x,y) dy$ Expected Value Rule $E[g(X,Y)] = \sum_x \sum_y g(x,y) p_{X,Y}(x,y)$ $E[g(X,Y)] = \int E[g(x,y) \mid Y=y] f_Y(y) dy$ $E[g(X,Y) \mid Y=y] = \int g(x,y) f_{X Y}(x \mid y) dy$ CDF $F_{X,Y}(x,y) = P(X \leq x, Y \leq y)$ $\phantom{F_{X,Y}(x,y)} = \int_{-\infty}^y \int_{-\infty}^x f_{X,Y}(s,t) ds dt$	Expected Value $E[g(x)] = \sum_x g(x) p_X(x)$ $E[g(x)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$ Linearity of Expectations $E[aX + b] = aE[X] + b$ Total Expectation Th. $E[X] = \sum_y p_Y(y) E[X \mid Y=y]$ $E[X] = \int_{-\infty}^{\infty} f_Y(y) E[X \mid Y=y] dy$ $E[X] = \sum_i P(A_i) E[X \mid A_i]$ Cond. Expectation $E[g(x) \mid Y=y] = \sum_x g(x) p_{X Y}(x \mid y)$ Iterated Expectation $E[E[X \mid Y]] = E[X]$
	DISCRETE DISTROS Bernouilli $P(X=1) = p$ $E[X] = p$ $Var(X) = p(1-p)$ Uniform $p_X(x) = \frac{1}{b-a+1}$ $E[X] = \frac{a+b}{2}$ $Var(X) = \frac{(b-a+1)^2 - 1}{12}$ Binomial <i>k successes in n trials</i> $p_X(k) = \binom{n}{k} p^k (1-p)^{n-k}$ $E[X] = np$ $Var(X) = np(1-p)$ Geometric <i>number of trials until success</i> $p_X(k) = (1-p)^{k-1} p$ $E[X] = \frac{1}{p}$ $Var(X) = \frac{1-p}{p^2}$ Poisson <i>how many occurrences k in τ given rate λ</i> $P(k, \tau) = \frac{(\lambda \tau)^k e^{-\lambda \tau}}{k!}$ $E[N_\tau] = \lambda \tau$ $Var(N_\tau) = \lambda \tau$	NORMALS $f_X(x) = \frac{1}{\sigma \sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$ $E[X] = \mu$ $Var(X) = \sigma^2$ Linear Functions $Y = aX + b$ with $X \sim N(\mu, \sigma^2)$ $Y = N(a\mu + b, a^2 \sigma^2)$ Indie Sum $Z \sim N(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2)$ Tables $\Phi(-2) = P(Y \leq -2)$ $ = 1 - P(Y \leq 2) = 1 - \Phi(2)$ Standardising $X \sim N(\mu, \sigma^2)$ and $Y \sim N(0, 1)$ $Y = \frac{X - \mu}{\sigma}$ $X = \mu + \sigma Y$	VARIANCE $Var(x) = E\left[(x - \mu)^2\right]$ and $\sigma = \sqrt{Var(X)}$ Properties $Var(aX + b) = a^2 Var(X)$ $Var(X) = E[X^2] - (E[X])^2$ Dependent Sum $Var(X_1 + \dots + X_n) = \sum_{i=1}^n Var(X_i) + \sum_{\{i,j\}: i \neq j} Cov(X_i, X_j)$ Independent Sum $Var(X + Y) = Var(X) + Var(Y)$	INDEPENDENCE If Indie $E[XY] = E[X]E[Y]$ $Var(X + Y) = Var(X) + Var(Y)$ $p_{X,Y Z}(x,y \mid z) = p_{X Z}(x \mid z) p_{Y Z}(y \mid z)$ $f_{X,Y}(x,y) = f_X(x) f_Y(y)$ $f_{X Y}(x \mid y) = f_X(x)$ $Cov(X,Y) = 0$ MIXED RV $X = Y$ (discrete) w.p p $ Z$ (continuous) w.p (1-p) $F_X(x) = pF_Y(x) + (1-p)F_Z(x)$ $E[X] = pE[Y] + (1-p)E[Z]$

<p>DERIVED DISTROS</p> <p>PMF function of discrete RV</p> $p_Y(y) = P(g(x) = y) = \sum_{x:g(x)=y} p_X(x)$ <p>Linear Functions</p> $Y = aX + b$ $p_Y(y) = p_X\left(\frac{y-b}{a}\right)$ $f_Y(y) = \frac{1}{ a } f_X\left(\frac{y-b}{a}\right)$ <p>g is monotonic</p> $f_Y(y) = f_X(h(y)) \left \frac{dh}{dy}(y) \right $ <p>general case</p> <p>1) find CDF: $F_Y(y) = P(g(x) \leq y)$</p> <p>2) derive CDF for PDF</p>	<p>BERNOUILLI PROCESS</p> <p>requires indie, time homogen.</p> <p>Properties</p> $S = X_1 + \dots + X_n$ $P(S = K) = \binom{n}{k} p^k (1-p)^{n-k}$ $E[S] = np$ $Var(S) = np(1-p)$ <p>Time until 1st success</p> $T_1 = \min \{i: X_i = 1\}$ $P(T_1 = k) = (1-p)^{k-1} p$ <p>Time of kth arrival</p> $p_{Y_k}(t) = \binom{t-1}{k-1} p^k (1-p)^{t-k}$ $E[Y_k] = \frac{k}{p}$ $Var(Y_k) = \frac{k(1-p)}{p^2}$ <p>Merging</p> $Z_t = g(X_t, Y_t) \sim Ber(p+q-pq)$ $\Rightarrow \text{prob either or both have arrival at time } t$ <p>Splitting</p> <p>flip a coin with prob q</p> $A \sim Ber(qp)$ $B \sim Ber((1-q)p)$ <p>Δ these streams are not indie</p>	<p>POISSON PROCESS</p> <p>indie, time homogen. seq of exp</p> <p>λ: arrival rate</p> $P(k, \tau) = \frac{(\lambda\tau)^k e^{-\lambda\tau}}{k!}$ $E[N_\tau] = \lambda\tau$ $Var(N_\tau) = \lambda\tau$ $\lambda = E \frac{N_\tau}{\tau}$ <p>Time of kth arrival / Erlang</p> $f_{Y_k} = \frac{\lambda^k y^{k-1} e^{-\lambda y}}{(k-1)!}$ $= Erlang(k)$ $= Erlang\left(\frac{k}{2}\right) + Erlang\left(\frac{k}{2}\right)$ <p>Sum</p> <p>Δ must be indie</p> <p>M: $Poisson(\mu)$ N: $Poisson(v)$</p> <p>M+N: $Poisson(\mu+v)$</p> <p>Merging</p> <p>A: λ_A B: λ_B</p> $\lambda = \lambda_A + \lambda_B$ $P(k^{th} \text{ arrival is A}) = \frac{\lambda_A}{\lambda_A + \lambda_B}$ $P(k \text{ arrivals are A}) \text{ is Binomial}\left(\frac{\lambda_A}{\lambda_A + \lambda_B}\right)$ <p>Splitting</p> <p>flip a coin with prob q</p> <p>Δ these streams <u>are</u> indie</p> <p>A: $\lambda_A = \lambda q$</p> <p>B: $\lambda_B = \lambda(1-q)$</p> <p>Multiple Engine Example</p> <p>3 engines with death rate λ_e</p> <p>rate until 1st dies is $\lambda = 3\lambda_e$</p> <p>then rate until 2nd dies $\lambda = 2\lambda_e$</p> <p>Min</p> $P(\min \{X, Y, Z\} \geq t)$ $= P(X \geq t, Y \geq t, Z \geq t)$ $= e^{-3\lambda t}$ $\Rightarrow \text{have 3 merged Poissons and want to know first arrival}$ $\Rightarrow \min \{X, Y, Z\} \text{ is } Exp(3\lambda)$ $E[\min \{X, Y, Z\}] = \frac{1}{3\lambda}$ <p>Max</p> $P(\max \{T_1, T_2, T_3\} \leq t)$ $= P(T_1 \leq t) P(T_2 \leq t) P(T_3 \leq t)$ $= (1 - e^{-\lambda t})^3$ <p>then derive this to get PDF</p>
<p>CONVOLUTIONS</p> $Z = X + Y$ $p_Z(z) = \sum_x p_X(x) p_Y(z-x)$ $f_Z(z) = \int_{-\infty}^{\infty} f_X(x) f_Y(z-x) dx$		
<p>COVARIANCE</p> $Cov(X, Y) = E[(X - E[X])(Y - E[Y])]$ <p>Direction</p> $Cov(X, Y) > 0 \text{ same sign}$ <p>If Indie</p> $Cov(X, Y) = 0$ <p>Δ inverse not true</p> <p>Properties</p> $Cov(X, X) = Var(X)$ $Cov(X, Y) = E[XY] - E[X]E[Y]$ $Cov(aX + b, Y) = aCov(X, Y)$ $Cov(X, Y + Z) = Cov(X, Y) + Cov(X, Z)$	<p>INTER-ARRIVAL TIMES / R. INCIDENCE</p>	
<p>CORRELATION COEF.</p> $\rho(X, Y) = \frac{Cov(X, Y)}{\sigma_X \sigma_Y}$	<p>we arrive at t^*</p> <p>u, v are each $Exp(\lambda)$ away from t^*</p> <p>$\Rightarrow E[V - U]$ is twice the expectation of $Exp(\lambda)$</p>	
<p>FRESH START/MEMORYLESSNESS</p> <p>Exponential</p> $f_{X X>t}(x x > t) = t + f_X(x)$ <p>Bernouilli/Poisson</p> $P(A B) = P(A)$ <p>i.e. prob of two arrivals (A) after 1 arrival (B) = prob of 2 arrivals (A)</p>		

