Note 13 Section 3.2. The Exponential, Gamma, and Chi-Square Distributions (D) Exponential distribution. · Recall the approx. Poisson process, where O # of cachironces in @ # of cacurrences The a small interval disjoint intervals are of length h indep, is almost Bernoulli w/ param 2h. Under this assumption, Poisson random variable = [# of occurrences on a] given interval.] Now, we ask: How long do we have to wait until the first occurrence? Let W = [Waiting time until the first occurrence]

• Then W is a non-negative, continuous RV. To
find the distribution, we investigate CDF:
For w >0, F(w) = P(W ≤ w)
= 1- P(W >w)
= 1- P(no occurrence in [0,w])
= 1- e^{-Nw}.
Condecomputed using
Prison distribution!
• PDF is the derivative of CDF:
f(w) = F'(w) =
$$\lambda e^{-\Lambda w}$$
, $W \ge 0$.
DEF A RV X has an expanontial distribution with mean
 $\theta = 1/\lambda$ if it has PDF:
f(x) = $\frac{1}{\theta} e^{-X/\theta}$, $x \ge 0$.
POP If X has exploit, w/ mean θ , then
E(X) = θ , $Var(X) = \theta^{*}$, $M_X(t) = \frac{1}{1-\theta t}$.
Pf) $M_X(t) = \int_{0}^{\infty} e^{tx} f(x) dx = \int_{0}^{\infty} e^{tx} \cdot \frac{1}{\theta} e^{-X/\theta} dx$
= $\begin{bmatrix} -\frac{1}{\theta} \cdot \frac{1}{t-V_{\theta}} e^{t} = -\frac{1}{1-\theta t}$.

Them $M'(t) = \frac{\theta}{(1-\theta t)^2} , \qquad M''(t) = \frac{2\theta^2}{(1-\theta t)^3} ,$ and so, $E(\chi) = \theta$, $Var(\chi) = (2\theta^2) - \theta^2 = \theta^2$. Ex Suppose a certain type of light bulk has an exp. dist. w/ mean life of 500 hours. Let X: life time of a light bulb. Then find: (a) P(X < 750), (b) P(X > 900 | X > 300). <u>Sol</u> (a) $P(X < 750) = \int_{1}^{750} \frac{1}{500} e^{-x/500} dx$ $= \left[e^{-x/500} \right]^{750} = 1 - e^{-1.5}$ or simply use the fact P(X(750) = F(750))(b) We have $P(X > 900 | X > 300) = \frac{P(\{X > 900\} \cap \{X > 300\})}{P(X > 300)}$ Since X>900 always implies X>300, $= \frac{P(\chi > 900)}{P(\chi > 300)}$ = $\frac{e^{-900/500}}{e^{-300/500}} = e^{-6/5}$ This coincides with $P(X > 6\infty) = e^{-6\infty/500}$. Note

In general, if X hus exp. dist. and
$$x, y > 0$$
,
 $P(X > x+y) | X > x) = P(X > y)$.
I.e. given that $X > x$, the "extra waiting time
 $X - x$ " behaves the same as exp. dist. of
the same mean. This is called "no-memory" property.
(2) Gamma distribution, special case.
(2) How long does it takes until the att accurrence?
EX
 $V = [waiting time for 4th occurrence]$.
Then for $w > 0$, its PDF is
 $f(w) \approx \frac{P(w \le W \le w + h)}{h}$
 $h small = \frac{1}{h} P(ath occurrence is in [w, wth])$
 $\approx \frac{1}{h} P(\{ath occurrence is in [w, wth])$
 $\sum_{in [w, wth]} \sum_{in [w, wth]} \sum_{i$

$$= \frac{1}{h} \cdot \frac{(\lambda w)^{\alpha-1}}{(\alpha-1)!} e^{-\lambda w} \cdot \frac{(\lambda h)'}{1!} e^{-\lambda h}$$

$$\approx \frac{\lambda^{\alpha} w^{\alpha-1}}{(\alpha-1)!} e^{-\lambda w}$$

$$= \frac{w^{\alpha-1}}{(\alpha-1)! \theta^{\alpha}} e^{-w/\theta} \quad (\theta = \frac{1}{\lambda})$$
As $h \rightarrow 0^{+}$, this becomes an identity. We say ∇V
has a gamma distribution.
(B) Gamma distribution, general case
• The above formula allows a generalization.
DEF The gamma function is defined by
 $\Gamma(t) = \int_{0}^{\infty} y^{t-1} e^{-y} dy, \quad t > 0.$
PROP (a) $\Gamma(1) = 1,$
(b) $\Gamma(t) = (t-1)\Gamma(t-1), \text{ for } t > 1.$
(c) $\Gamma(n) = (n-1)!, \text{ for } n = 1, 2, 3, \cdots.$
I.e., Γ generalizes factorial.
• Using this, we define:
DEF X has a gamma distribution if its PDF is
 $f(\alpha) = \frac{1}{\Gamma(\alpha) \theta^{\alpha}} \alpha^{\alpha-1} e^{-\alpha/\theta}, \quad \alpha > 0,$
 $\theta > 0,$

Prop If X has gamma dist., then

$$E(X) = d\theta$$
, $Var(X) = a\theta^2$, $M(t) = \frac{1}{(1-\theta t)^N}$.
(A) Chi-Square Distribution
Q What is the use of gamma dist of non-integer a?
A Same other distributions, arising in different
contexts, reduce to gamma dist.
DEF X has chi-square distribution w/ r degree of
freedom if it has a gamma dist. w/ $\theta = 2$ and $\alpha = \frac{1}{5}$.
I.e.,
 $f(x) = \frac{1}{\Gamma(r/2)} 2^{r/2} x^{\frac{K}{2}+1} e^{-\chi/2}$, $\chi > 0$.
This is often abbeviated by saying:
 $X \sim \chi^2(r)$.
 $F(X) = 2 \cdot \frac{K}{2} = r$, $Var(X) = 2^2 \cdot \frac{K}{2} = 2r$.
 $\chi^2_{\alpha}(r) := [100(1-\alpha)^{\frac{1}{10}} percentile].$