

Introduction to probability

Stat 134

FAII 2005 Berkeley

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Follows Jim Pitman's book: Probability Section 3.5

### The Poisson ( $\mu$ ) Distribution

The Poisson distribution with parameter  $\mu$  or Poisson( $\mu$ ) distribution is the distribution of probabilities P<sub>u</sub>(k) over {0,1,2,...} defined by:

$$P(k) = e^{-\mu} \frac{\mu^{k}}{k!}$$

## **Poisson Distribution**

•Recall that  $Poisson(\mu)$  is a good approximation for N = Bin(n,p) where

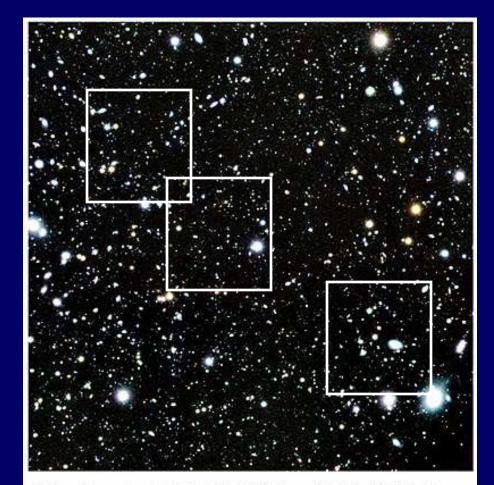
- •n is large
- •p is small and
- µ = np.

## **Poisson Distribution**

#### Example:

#### N<sub>galaxies</sub>:

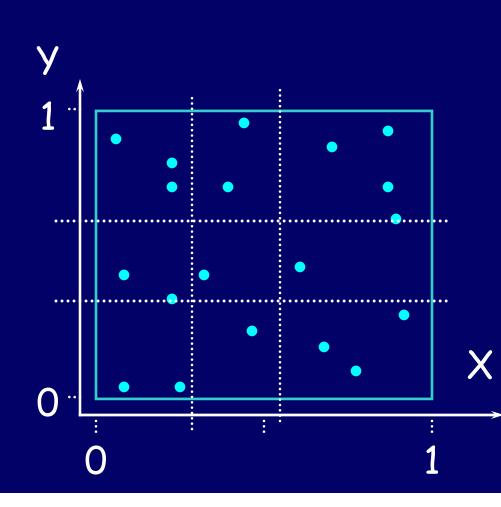
the number of supernova found in a square inch of the sky is distributed according to the Poisson distribution.



Colour-Composite of the Sky Field With Several High-Redshift Galaxies VLT YEPUN + FORS2

ESO PR Photo 25a/03 (21 August 2003 )

## **Poisson Distribution**



#### Example:

N<sub>drops</sub>: the number of rain drops falling on a given square inch of the roof in a given period of time has

a Poisson distribution.

Poisson Distribution:  $\mu$  and  $\sigma$ Since N = Poisson( $\mu$ ) ~ Bin(n, $\mu$ /n) we expect that: E(N) = lim (np) =  $\mu$ ;

SD(N) =  $\lim_{np=\mu, p\to 0, n\to\infty} \sqrt{\mu}$ .

 $np=\mu,p\rightarrow 0,n\rightarrow\infty$ 

Next we will verify it.

# Mean of the Poisson Distribution

#### By direct computation:

$$E[N] = \sum_{k=0}^{\infty} kP[N=k] = \sum_{k=1}^{\infty} ke^{-\mu} \frac{\mu^{k}}{k!}$$
$$= e^{-\mu} \mu \sum_{k=1}^{\infty} \frac{\mu^{k-1}}{(k-1)!} = e^{-\mu} \mu \sum_{k=0}^{\infty} \frac{\mu^{k}}{k!}$$
$$= e^{-\mu} \mu e^{\mu} = \mu.$$

# SD of the Poisson Distribution

$$E[N(N-1)] = \sum_{k=0}^{\infty} k(k-1)P[N=k] = \sum_{k=2}^{\infty} k(k-1)e^{-\mu}\frac{\mu^{k}}{k!}$$
$$= e^{-\mu}\mu^{2}\sum_{k=2}^{\infty}\frac{\mu^{k-2}}{(k-2)!} = e^{-\mu}\mu^{2}\sum_{k=0}^{\infty}\frac{\mu^{k}}{k!}$$
$$= e^{-\mu}\mu^{2}e^{\mu} = \mu^{2}.$$

So

$$E[N^{2}] = E[N(N-1)] + E[N] = \mu^{2} + \mu$$
$$Var[N] = E[N^{2}] - E[N]^{2} = (\mu^{2} + \mu) - \mu^{2} = \mu.$$
$$SD(N) = \sqrt{\mu}.$$

Sum of independent Poissions Consider two independent r.v.'s :  $N_1 \sim \text{Poisson}(\mu_1)$  and  $N_2 \sim \text{Poisson}(\mu_2)$ . •We approximately have:  $N_1 \sim Bin(\mu_1/p,p)$  and  $N_2 \sim Bin(\mu_2/p,p)$ , where p is small. • So if we pretend that  $\mu_1/p_{\mu_2}/p$  are integers then:

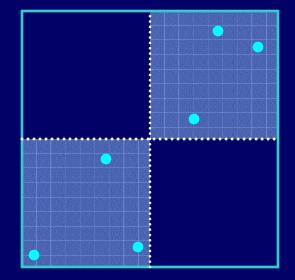
 $N_1 + N_2 \sim Bin((\mu_1 + \mu_2)/p,p) \sim Poisson(\mu_1 + \mu_2)$ 

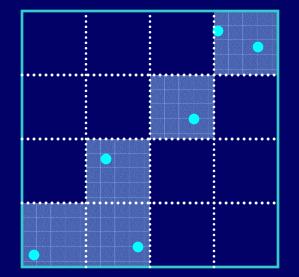
Sum of independent Poissions •<u>Claim</u>: if  $N_i \sim Poisson(\mu_i)$  are independent then  $N_1 + N_2 + \dots + N_r \sim Poisson(\mu_1 + \dots + \mu_r)$ •Proof (for r=2):  $P[N_1 + N_2 = k] = \sum_{i=1}^{n} P[N_1 = j]P[N_2 = k - j]$ i=0 $= \sum_{j=0}^{k} e^{-\mu_1} \frac{\mu_1^j}{j!} e^{-\mu_2} \frac{\mu_2^{k-j}}{(k-j)!}$  $= \frac{e^{-(\mu_1 + \mu_2)}}{k!} \sum_{j=0}^k {k \choose j} \mu_1^j \mu_2^{k-j}$  $(\mu_1 + \mu_2)^k e^{-(\mu_1 + \mu_2)}$ k!

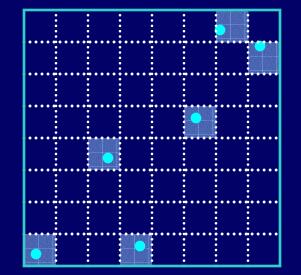
<u>The Poisson Scatter Theorem:</u> Considers particles hitting the square where:

- Different particles hit different points and
- If we partition the square into n equi-area squares then each square is hit with the same probability p<sub>n</sub> independently of the hits of all other squares

- The Poisson Scatter Theorem: states that under these two assumptions there exists a number  $\lambda > 0$  s.t:
- For each set B, the number N(B) of hits in B satisfies N(B)  $\sim Poisson(\lambda \times Area(B))$
- For disjoint sets B<sub>1</sub>,...,B<sub>r</sub>, the numbers of hits N(B<sub>1</sub>), ...,N(B<sub>r</sub>) are independent.
- The process defined by the theorem is called Poisson scatter with intensity  $\lambda$







## Poisson Scatter Thinning

- <u>Claim</u>: Suppose that in a Poisson scatter with intensity  $\lambda$
- each point is kept with prob. p and erased with probability 1-p
- independently of its position and the erasure of all other points.
- Then the scatter of points kept is a Poisson scatter with intensity p  $\lambda.$