## Next two weeks: projects

- Short (20-30 minute) videoclips of lectures will be posted on the website (the last column). Today's material is here: <a href="Erlang and Gamma">Erlang and Gamma</a>
- For each of the next 4 lectures, there will be a project assignment to be solved in class
- I will not be present in class, but I recommend that you meet in person two times a week during regular hours
- All 4 projects will be submitted as one document from each of the 4 groups
- The projects are worth 20% of the grade, the midterm -30%, the final exam -50%

Constant vale (POTSSON) process discrete events happen at rate [ Expected number of events in time oc The actual number of events Na 15 a Poisson distributed discrete random variable  $P(N=n)=\frac{r_{c}}{h_{1}}e^{-r_{c}}$ Why Poisson? Divide X into many tiny intervals of Length DX Prob(N=n)= (L)pn(1-p)L·n P= Pox L= x/ox  $E(N_{E}) = \rho L = \Gamma x$ Poisson

### Constant rate (AKA Poisson) processes

- Let's assume that proteins are produced by ribosomes in the cell at a rate r per second.
- The expected number of proteins produced in x seconds is  $r \cdot x$ .
- The actual number of proteins N<sub>x</sub> is a discrete random variable following a Poisson distribution with mean r·x:

$$P_N(N_x=n)=\exp(-r\cdot x)(r\cdot x)^n/n!$$
  $E(N_x)=rx$ 

- Why Discrete Poisson Distribution?
  - Divide time into many tiny intervals of length  $\Delta x << 1/r$
  - The probability of success (protein production)
     per internal is small: p\_success=r∆x <<1,</li>
  - The number of intervals is large:  $n = x/\Delta x >> 1$
  - Mean is constant:  $r=E(N_x)=p_success \cdot n = r\Delta x \cdot x/\Delta x = r \cdot x$
  - In the limit  $\Delta x << x$ , p\_success is small and n is large, thus Binomial distribution → Poisson distribution

# **Exponential Distribution Definition**

Exponential random variable X describes interval between two successes of a constant rate (Poisson) random process with success rate r per unit interval.

The probability density function of *X* is:

$$f(x) = re^{-rx}$$
 for  $0 \le x < \infty$ 

Closely related to the discrete geometric distribution

$$f(x) = p(1-p)^{x-1} = p e^{(x-1) \ln(1-p)} \approx pe^{-px}$$
 for small p

o summarite constant rate processes: time I - rate per unit of length N(x) - disrese number of events Toisson: P(N(x)=h)= (r,x)n - r.x

Nine x

(r,x)n - r.x Time interval X between 5400essive events 15

continuously distributed vandom variable

Its PDF if  $f(x) = e^{-rx}$ 

# What is the interval X between two successes of a constant rate process?

- X is a continuous random variable
- CCDF:  $P_X(X>x) = P_N(N_X=0) = exp(-r \cdot x)$ .
  - Remember:  $P_N(N_x=n)=exp(-r\cdot x) (r\cdot x)^n/n!$
- PDF:  $f_X(x) = -dCCDF_X(x)/dx = r \cdot exp(-r \cdot x)$
- We started with a discrete Poisson distribution where time x was a parameter
- We ended up with a continuous exponential distribution

# Exponential Mean & Variance

If the random variable *X* has an exponential distribution with rate r,

$$\mu = E(X) = \frac{1}{r}$$
 and  $\sigma^2 = V(X) = \frac{1}{r^2}$  (4-15)

#### Note that, for the:

- Poisson distribution: mean= variance
- Exponential distribution: mean = standard deviation = variance<sup>0.5</sup>

#### **Biochemical Reaction Time**

 The time x (in minutes) until all enzymes in a cell catalyze a biochemical reaction and generate a product is approximated by this CCDF:

$$F_{>}(x) = e^{-2x}$$
 for  $0 \le x$ 

Here the rate of this process is r=2 min<sup>-1</sup> and 1/r=0.5 min is the average time between successive products of these enzymes

What is the PDF?

$$f(x) = -\frac{dF_{>}(x)}{dx} = -\frac{d}{dx}e^{-2x} = 2e^{-2x} \text{ for } 0 \le x$$

 What proportion of reactions will not generate another product within 0.5 minutes of the previous product?

$$P(X > 0.5) = F_{>}(0.5) = e^{-2*0.5} = 0.37$$

# We observed our cell for 1 minute and no product has been generated: The product is "overdue"

What is the probability that a product will not appear during the next 0.5 minutes?

$$F_{>}(x) = e^{-2x}$$
  
 $F_{>}(0.5) \approx 0.37$   
 $F_{>}(1.5) \approx 0.05$   
 $F_{>}(1.0) \approx 0.13$ 

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Memoryless property of the exponential P(X>t+s|X>s) = P(X>t) $P(X>t+s\mid X>s) = \frac{P(X>t+s, X>s)}{P(X>s)} =$  $=\frac{e\times p(-\Upsilon(t+s))}{e\times p(-\chi s)}=\frac{e\times p(-\Upsilon(t))}{e\times p(-\chi s)}$  $= \mathcal{P}(X > t)$ Exponential is the only memoryless distribution

#### Matlab exercise:

- Generate a sample of 100,000 variables from Exponential distribution with r = 0.1
- Calculate mean and compare it to 1/r
- Calculate standard deviation and compare it to 1/r
- Plot semilog-y plots of PDFs and CCDFs.
- Hint: read the help page (better yet documentation webpage) for random('Exponential'...) one of their parameters is different than r

# Matlab exercise: Exponential

```
Stats=100000; r=0.1;

    r2=random('Exponential', 1./r, Stats,1);

disp([mean(r2),1./r]); disp([std(r2),1./r]);
step=1; [a,b]=hist(r2,0:step:max(r2));
pdf_e=a./sum(a)./step;
subplot(1,2,1); semilogy(b,pdf_e,'rd-');
• x=0:0.01:max(r2);
for m=1:length(x);
    ccdf_e(m)=sum(r2>x(m))./Stats;

    end;
```

subplot(1,2,2); semilogy(x,ccdf\_e,'ko-');

### **Erlang Distribution**

- The Erlang distribution is a generalization of the exponential distribution.
- The exponential distribution models the time interval to the 1<sup>st</sup> event, while the
- Erlang distribution models the time interval to the k<sup>th</sup> event, i.e., a sum of k exponentially distributed variables.
- The exponential, as well as Erlang distributions, is based on the constant rate (or Poisson) process.

Constant vale (POTSSON) process Events happen independently
from each other at
constant rate= [: ENa]=Fix Follows Erlang distribution  $f(X>x)=\sum P(N_x=n)=$  $= \sum_{n=1}^{\infty} \frac{(rx)^n n = 0}{n!}$ 

### **Erlang Distribution**

Generalizes the Exponential Distribution: waiting time until k's events (constant rate process with rate=r)

$$P(X > x) = \sum_{m=0}^{k-1} \frac{e^{-rx}(rx)^m}{m!} = 1 - F(x)$$

Differentiating F(x) we find that all terms in the sum except the last one cancel each other:

$$f(x) = \frac{r^k x^{k-1} e^{-rx}}{(k-1)!}$$
 for  $x > 0$  and  $k = 1, 2, 3, ...$ 

#### Gamma Distribution

The random variable *X* with a probability density function:

$$f(x) = \frac{r^k x^{k-1} e^{-rx}}{\Gamma(k)}, \text{ for } x > 0$$
 (4-18)

has a gamma random distribution with parameters r > 0 and k > 0. If k is a positive integer, then X has an Erlang distribution.



$$f(x) = \frac{r^k x^{k-1} e^{-rx}}{\Gamma(k)}, \text{ for } x > 0$$

$$\int_{0}^{+\infty} f(x) dx = 1, \text{ Hence}$$

$$\Gamma(k) = \int_{0}^{+\infty} r^{k} x^{k-1} e^{-rx} dx = \int_{0}^{+\infty} y^{k-1} e^{-y} dy$$

Comparing with Erlang distribution for integer k one gets

$$\Gamma(k) = (k-1)!$$

#### **Gamma Function**

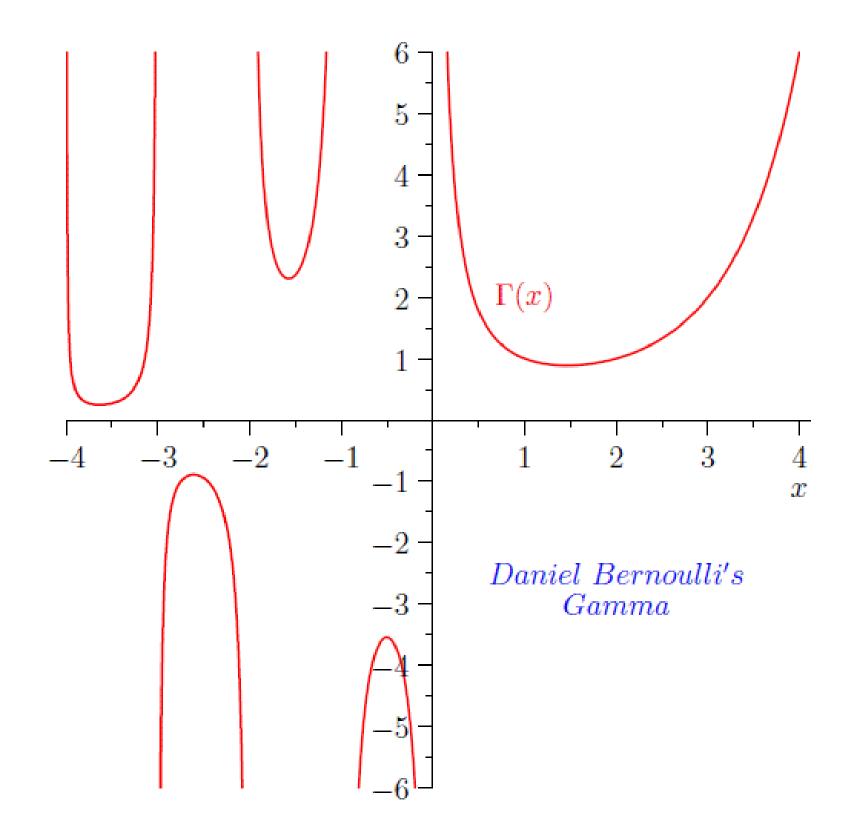
The gamma function is the generalization of the factorial function for r > 0, not just non-negative integers.

$$\Gamma(k) = \int_{0}^{\infty} y^{k-1} e^{-y} dy$$
, for  $r > 0$  (4-17)

Properties of the gamma function

$$\Gamma(1) = 1$$

$$\Gamma(k) = (k-1)\Gamma(k-1)$$
 recursive property
$$\Gamma(k) = (k-1)!$$
 factorial function
$$\Gamma(1/2) = \pi^{1/2} = 1.77$$
 interesting fact





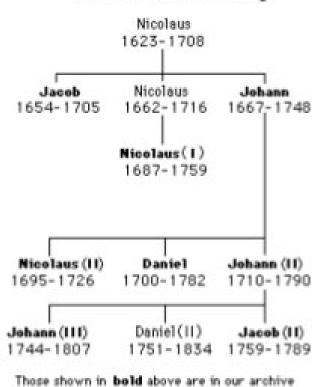
# Bernoulli FAMILY Thats

# BERNOULLI FAMILY

#### **SOLO HERMELIN**

http://www.solohermelin.com

#### The Bernoulli family







Jacob 1654-1705



Johann 1667-1748



Nicolaus II 1695-1720



Daniel 1700-1782



Johann II 1710-1790



Johann III 1744-1807



Jacob II 1759-1789

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Samma f

function

# Mean & Variance of the Erlang and Gamma

• If X is an Erlang (or more generally Gamma) random variable with parameters r and k,  $\mu = E(X) = k/r$  and  $\sigma^2 = V(X) = k/r^2$  (4-19)

• Generalization of exponential results:  $\mu = E(X) = 1/r$  and  $\sigma^2 = V(X) = 1/r^2$  or Negative binomial results:  $\mu = E(X) = k/p$  and  $\sigma^2 = V(X) = k(1-p) / p^2$ 

#### Matlab exercise:

- Generate a sample of 100,000 variables with "Harry Potter" Gamma distribution with r = 0.1 and k=9 ¾ (9.75)
- Calculate mean and compare it to k/r (Gamma)
- Calculate standard deviation and compare it to sqrt(k)/r (Gamma)
- Plot semilog-y plots of PDFs and CCDFs.
- Hint: read the help page (better yet documentation webpage) for random('Gamma'...): one of their parameters is different than r

#### Matlab exercise: Gamma

```
    Stats=100000; r=0.1; k=9.75;

r2=random('Gamma', k,1./r, Stats,1);
disp([mean(r2),k./r]);
 disp([std(r2),sqrt(k)./r]);
step=0.1; [a,b]=hist(r2,0:step:max(r2));
 pdf_g=a./sum(a)./step;
figure;
 subplot(1,2,1); semilogy(b,pdf_g,'ko-'); hold on;
x=0:0.01:max(r2); clear cdf_g;
for m=1:length(x);
    cdf_g(m)=sum(r2>x(m))./Stats;
  end;
  subplot(1,2,2); semilogy(x,cdf g,'rd-');
```

