

**Raoul LePage**

**Professor**

**STATISTICS AND PROBABILITY**

**[www.stt.msu.edu/~lepage](http://www.stt.msu.edu/~lepage)**

**click on STT200\_Sp09**

**Lecture Outline, 3-27-09 and 3-30-09. See pp. 436-444.**

# RANDOM VARIABLE

## Chapter 16

**boats sold** probability

2	0.2
3	0.2
4	0.3
5	0.1
6	0.1
7	0.05
8	<u>0.05</u>
total	1

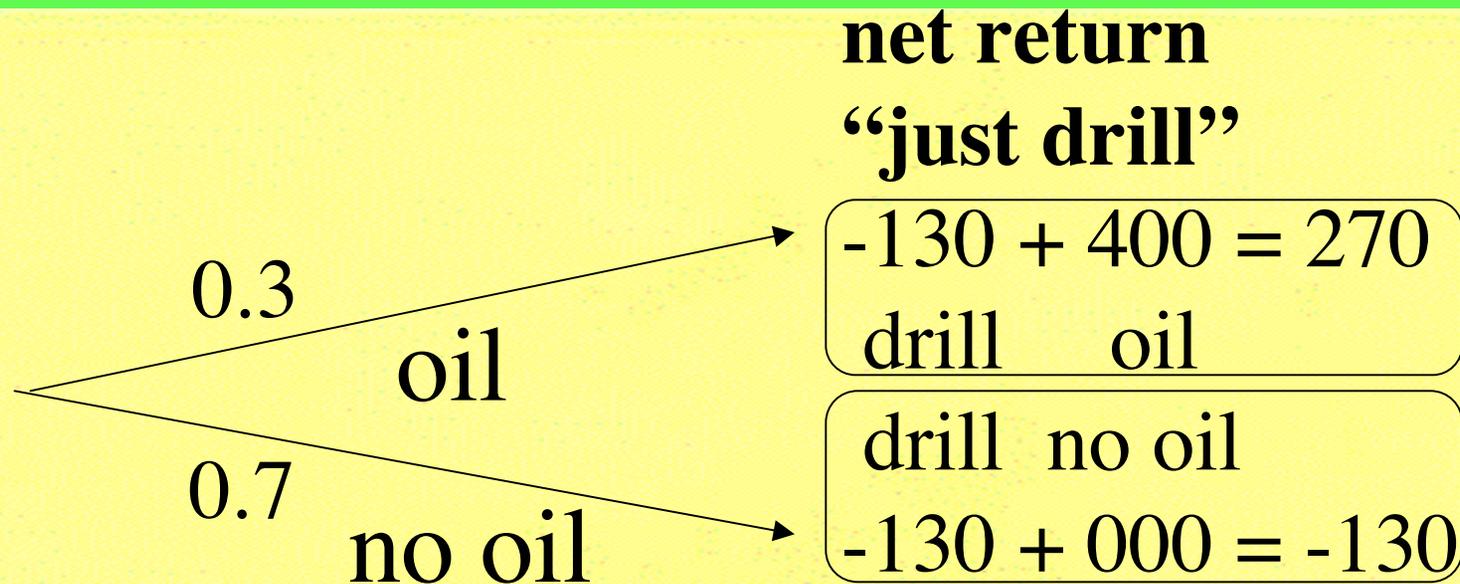
$P(\text{fewer than } 3.7) = .4$

$P(4 \text{ to } 7) = .55$

# OIL DRILLING EXAMPLE

$$P(\text{oil}) = 0.3$$

Cost to drill 130  
Reward for oil 400



A random variable is just a **numerical function** over the outcomes of a probability experiment.<sup>3</sup>

# EXPECTATION

## Definition of $E X$

$E X =$  sum of value times probability  $\times p(x)$ .

## Key properties

$$E(a X + b) = a E(X) + b$$

$$E(X + Y) = E(X) + E(Y) \text{ (always, if such exist)}$$

a.  $E(\text{sum of 13 dice}) = 13 E(\text{one die}) = 13(3.5)$ .

b.  $E(0.82 \text{ Ford US} + \text{Ford Germany} - 20M)$   
 $= 0.82 E(\text{Ford US}) + E(\text{Ford Germany}) - 20M$

regardless of any possible dependence.

# total of 2 dice

	<u>probability</u>	<u>product</u>	(3-15) of text
2	1/36	2/36	<b>E ( total ) is just twice the 3.5 avg for one die</b>
3	2/36	6/36	
4	3/36	12/36	
5	4/36	20/36	
6	5/36	30/36	
7	6/36	42/36	
8	5/36	40/36	
9	4/36	36/36	
10	3/36	30/36	
11	2/36	22/36	
12	<u>1/36</u>	<u>12/36</u>	
sum	1	$252/36 = 7$	

(3-17 of text)

**boats/month**

	<u>probability</u>	<u>product</u>
2	0.2	0.4
3	0.2	0.6
4	0.3	1.2
5	0.1	0.5
6	0.1	0.6
7	0.05	0.35
8	0.05	0.4
total	1	4.05

**we avg  
4.05 boats  
per month**

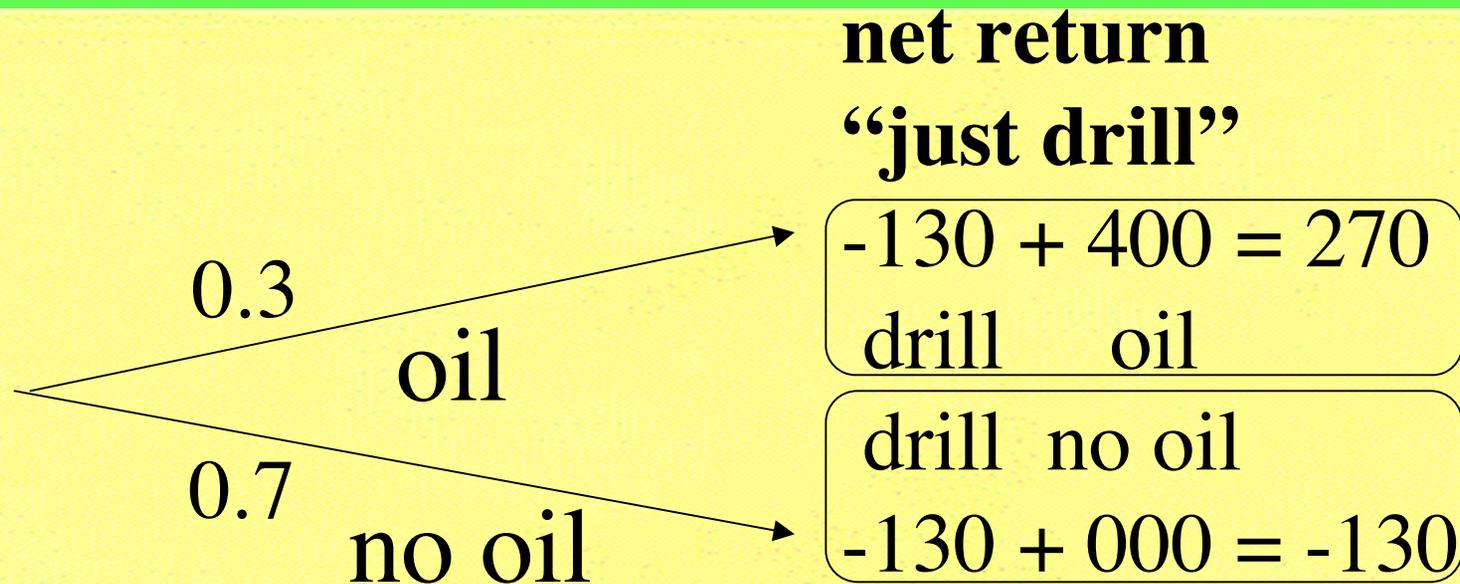
**E(number of boats this month)**



# OIL DRILLING EXAMPLE

$$P(\text{oil}) = 0.3$$

Cost to drill 130  
Reward for oil 400



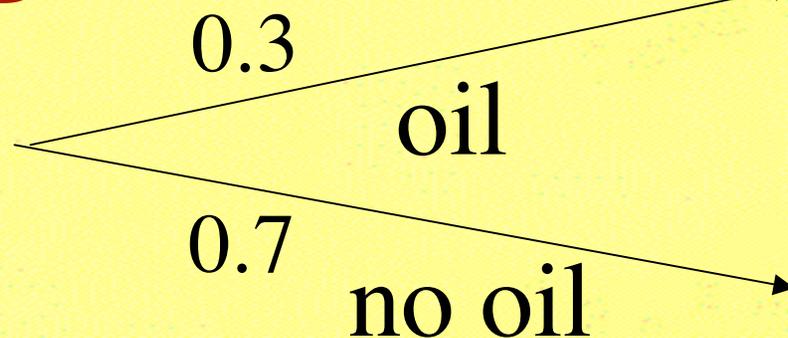
A random variable is just a **numerical function** over the outcomes of a probability experiment.<sup>7</sup>

# EXPECTATION IN THE OIL EXAMPLE

Expected return from policy “just drill” is the probability weighted average (NET) return

$$E(\text{NET}) = (0.3) (270) + (0.7) (-130) = 81 - 91 = -10.$$

**just drill**



net return from policy “just drill.”

$$-130 + 400 = 270$$

drill oil

drill no-oil

$$-130 + 0 = -130$$

$$E(X) = -10$$

# OIL EXAMPLE WITH A "TEST FOR OIL"

"costs"

TEST 20

DRILL 130

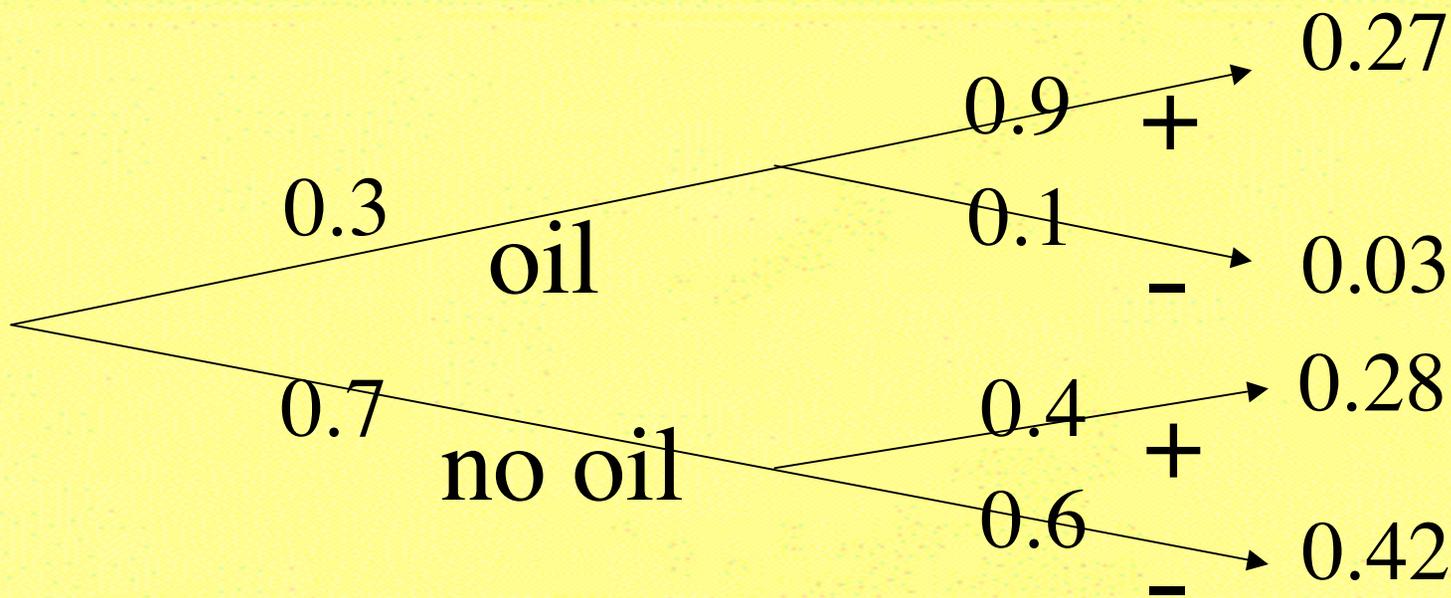
OIL 400

A test costing 20 is available.

This test has:

$$P(\text{test } + \mid \text{oil}) = 0.9$$

$$P(\text{test } + \mid \text{no-oil}) = 0.4.$$



Is it worth 20 to test first?

# EXPECTED RETURN IF WE "TEST FIRST"

	net return	prob	prod
oil+	$-20 - 130 + 400 = 250$	0.27	67.5
oil-	$-20 - 0 + 0 = -20$	.03	- 0.6
no oil+	$-20 - 130 + 0 = -150$	.28	- 42.0
no oil-	$-20 - 0 + 0 = -20$	.42	- 8.4
<b>drill only if the test is +</b>		total 1.00	<b>16.5</b>

$$E(\text{NET}) = .27 (250) - .03 (20) - .28 (150) - .42 (20) = 16.5 \text{ (for the "test first" policy).}$$

This average return is much preferred over the  $E(\text{NET}) = -10$  of the "just drill" policy.

# Variance and s.d. of boats/month

(3-17)  
of text

x	p(x)	x p(x)	x <sup>2</sup> p(x)	(x-4.05) <sup>2</sup> p(x)
2	0.2	0.4	0.8	0.8405
3	0.2	0.6	1.8	0.2205
4	0.3	1.2	4.8	0.0005
5	0.1	0.5	2.5	0.09025
6	0.1	0.6	3.6	0.38025
7	0.05	0.35	2.45	0.435125
8	0.05	0.4	3.2	0.780125
total	1.00	4.05	19.15	2.7475
quantity		E X	E X <sup>2</sup>	E (X - E X) <sup>2</sup>
terminology		mean	mean of squares	variance = mean of sq dev

$$\text{s.d.} = \text{root}(2.7474) = \text{root}(19.15 - 4.05^2) = 1.6576$$

# VARIANCE AND STANDARD DEVIATION

$$\text{Var}(X) \stackrel{\text{def}}{=} E (X - E X)^2 \stackrel{\text{comp}}{=} E (X^2) - (E X)^2$$

i.e.  $\text{Var}(X)$  is the expected square deviation of r.v.  $X$  from its own expectation.

**Caution:** The computing formula (right above), although perfectly accurate mathematically, is sensitive to rounding errors.

**Key properties:**

$$\text{Var}(a X + b) = a^2 \text{Var}(X) \text{ (b has no effect).}$$

$$\text{sd}(a X + b) = |a| \text{sd}(X).$$

$$\text{VAR}(X + Y) = \text{Var}(X) + \text{VAR}(Y) \text{ if } X \text{ ind of } Y.$$

# EXPECTATION AND INDEPENDENCE

**Random variables  $X, Y$  are INDEPENDENT if**

$p(x, y) = p(x) p(y)$  for all possible values  $x, y$ .

**If random variables  $X, Y$  are INDEPENDENT**

$E(X Y) = (E X) (E Y)$  echoing the above.

$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$ .

# PRICE RELATIVES

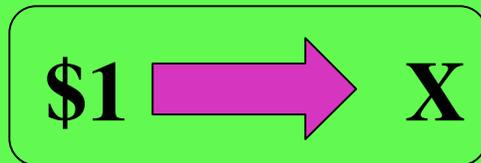
Venture one returns random variable  $X$  per \$1 investment. This  $X$  is termed the “price relative.” This random  $X$  may in turn be reinvested in venture two which returns random random variable  $Y$  per \$1 investment. The return from \$1 invested at the outset is the product random variable  $XY$ .

## EXPECTED RETURN

If INDEPENDENT,  $E( X Y ) = (E X) (E Y)$ .

# PARADOX OF GROWTH

**EXAMPLE:**



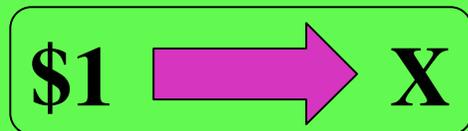
x	p(x)	x p(x)
0.8	0.3	0.24
1.2	0.5	0.60
1.5	0.2	<u>0.30</u>
		$E(X) = 1.14$

**WE AVERAGE 14% PER PERIOD**

**BUT YOU WILL NOT EARN 14%. Simply put, the average is not a reliable guide to real returns in the case of exponential growth.**

# EXPECTATION governs SUMS but sums are in the exponent

EXAMPLE:



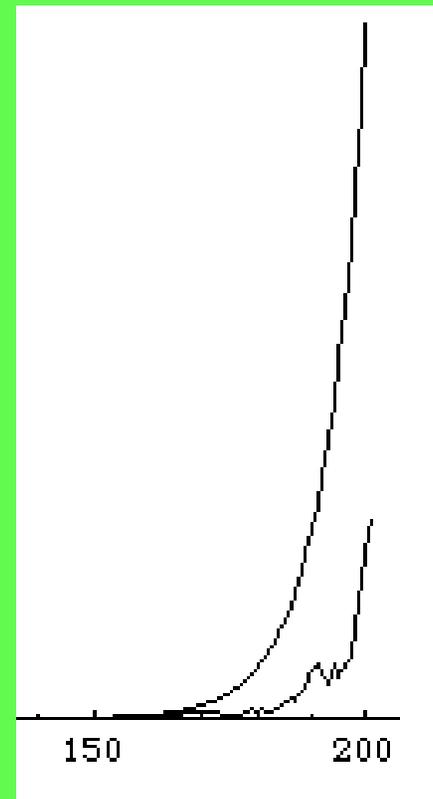
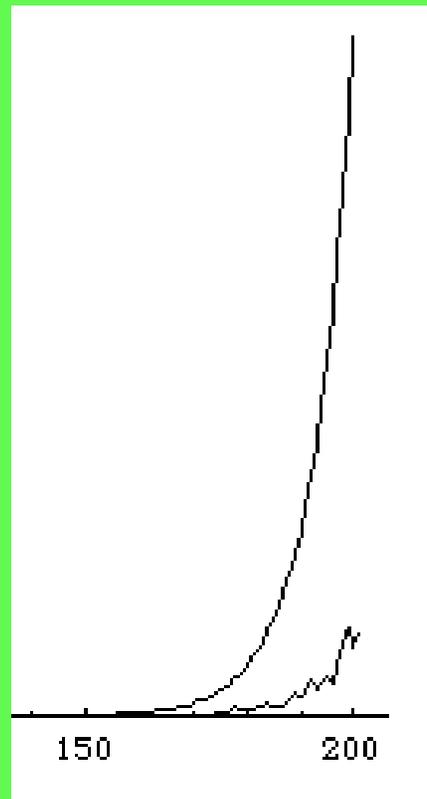
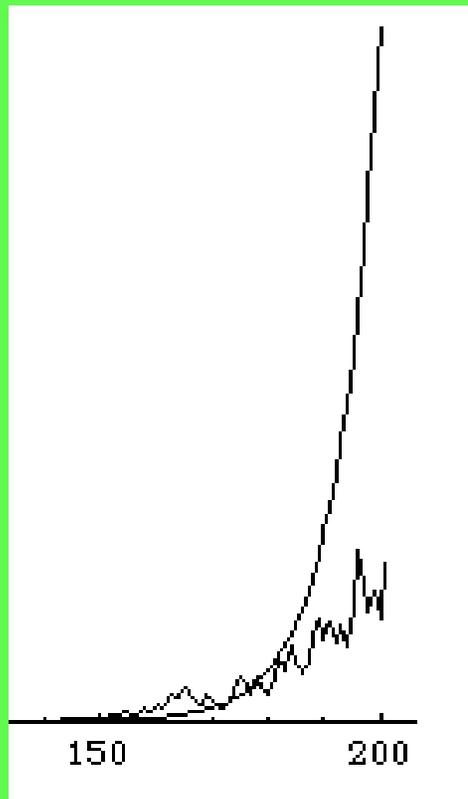
x	p(x)	$\text{Log}_e[x] p(x)$
0.8	0.3	-0.029073
1.2	0.5	0.039591
1.5	0.2	<u>0.035218</u>
		$E \text{Log}_e[X] = 0.105311$

$$e^{0.105311..} \leftarrow \text{---} \rightarrow \cong 1.11106..$$

With INDEPENDENT plays your RANDOM return will compound at 11.1% not 14%.

(more about this later in the course)

# COMPARING $1.14^n$ WITH THREE RANDOM EVOLUTIONS



you can see that 14% exceeds reality

# Poisson Distribution Governing Counts of Rare Events

$$p(x) = e^{-\text{mean}} \text{mean}^x / x!$$

for  $x = 0, 1, 2, \dots$  ad infinitum

# Poisson

**first best thing:**

**THE FIRST BEST THING ABOUT THE POISSON IS THAT THE MEAN ALONE TELLS US THE ENTIRE DISTRIBUTION!**

**note: Poisson sd =  $\sqrt{\text{mean}}$**

**Poisson Cookies**  
**400 raisins**  
**144 COOKIES**  
**mix well**

**$E X = 400/144 \sim 2.78$  raisins per cookie**  
 **$sd = \text{root}(\text{mean}) = 1.67$**   
**(for Poisson)**

# Poisson Cookies

e.g.  $X$  = number of raisins in MY cookie. Batter has 400 raisins and makes 144 cookies.

$$E X = 400/144 \sim 2.78 \text{ per cookie}$$

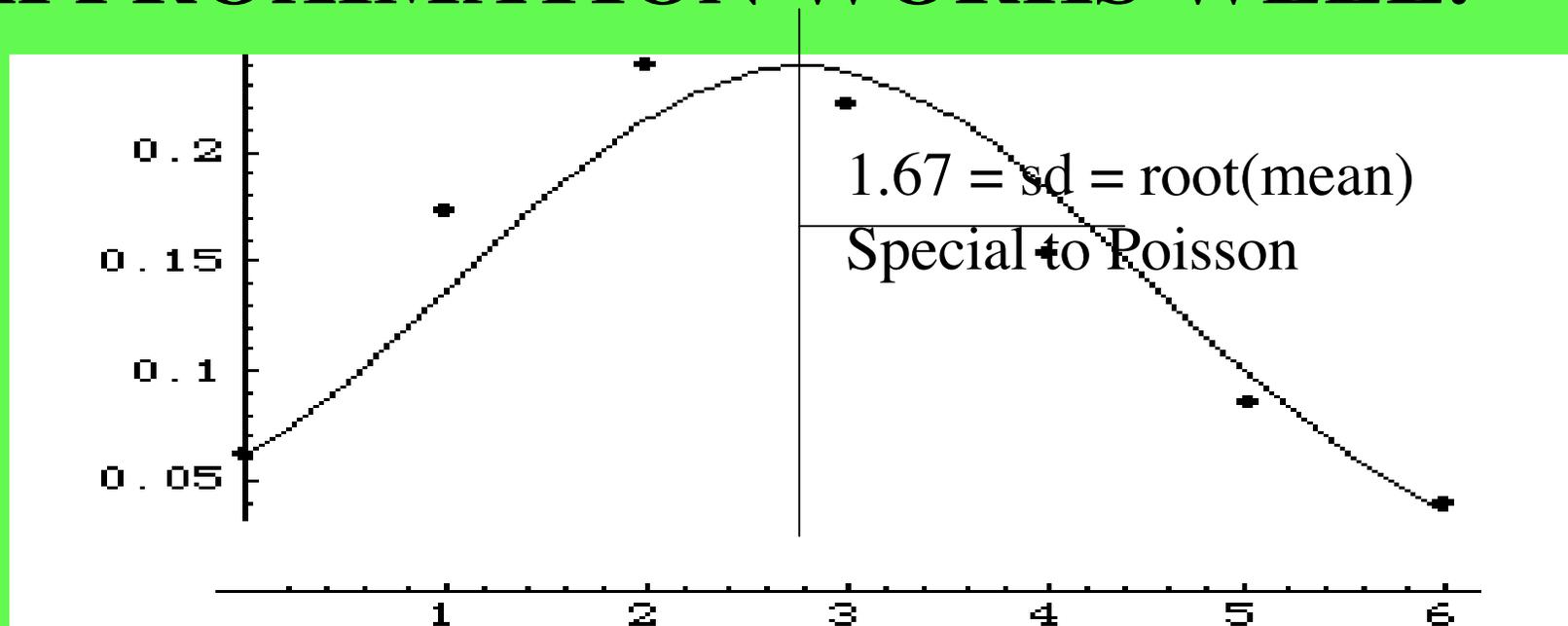
$$p(x) = e^{-\text{mean}} \text{mean}^x / x!$$

e.g.  $p(2) = e^{-2.78} 2.78^2 / 2! \sim 0.24$   
(around 24% of cookies have 2 raisins)

# Poisson

## second best thing

THE SECOND BEST THING ABOUT THE POISSON IS THAT FOR A MEAN AS SMALL AS 3 THE NORMAL APPROXIMATION WORKS WELL.



mean 2.78

# Poisson at cards

e.g.  $X$  = number of times ace of spades turns up in 104 independent tries (i.e. from full deck)

$X \sim$  Poisson with mean 2

$$p(x) = e^{-\text{mean}} \text{mean}^x / x!, \quad x=0 \dots$$

$$p(3) = e^{-2} 2^3 / 3! \sim 0.182205$$

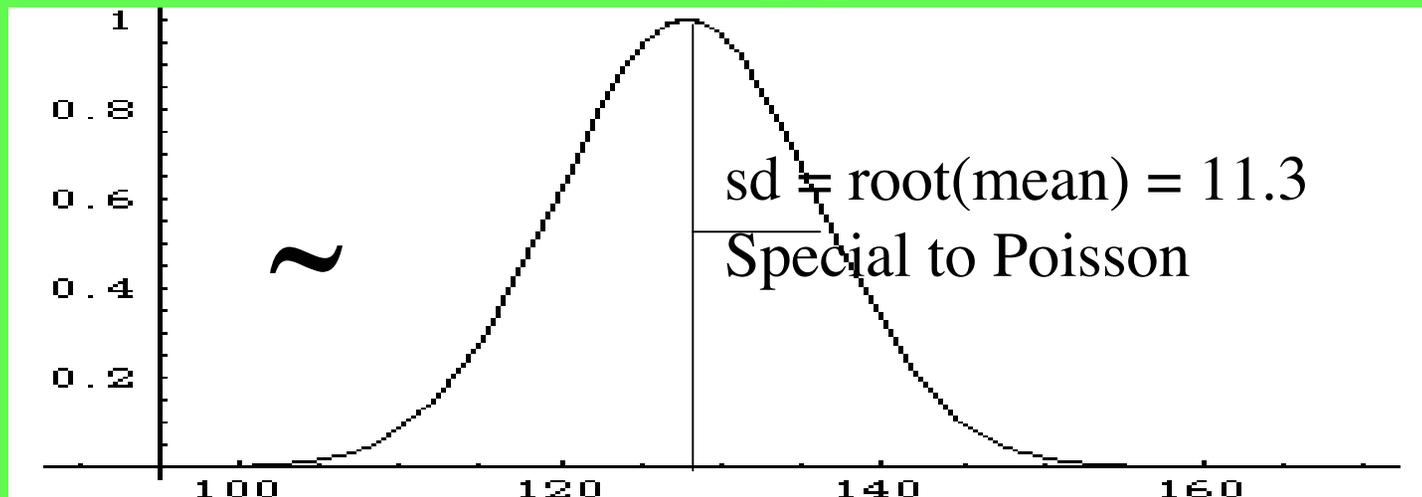
# Poisson in Risk

## AVERAGE 127.8

### ACCIDENTS PER MO.

$E X = 127.8$  accidents

If Poisson then  $sd = \text{root}(127.8) = 11.3049$  and the approx dist is:



mean 127.8 accidents

# Binomial

$X = \#$  of successes,  $n$  independent tries

e.g.  $X =$  number of times ace of spades turns up in 104 deals of 1 card from a shuffled full deck.

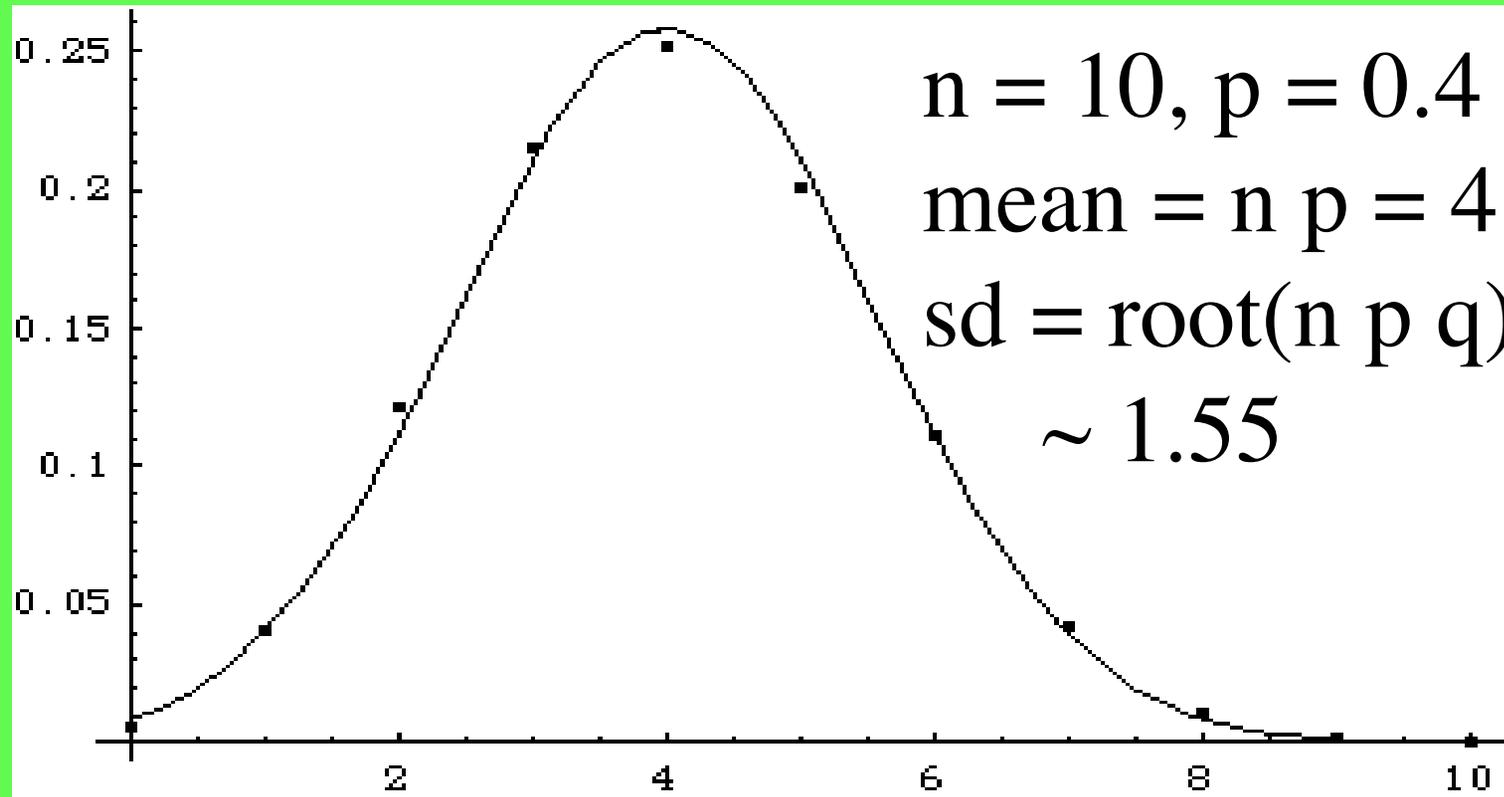
Binomial ( $n=104$ ,  $p = 1/52$ )

$$p(x) = nC_x * p^x q^{n-x}, \quad n = 0 \text{ to } n.$$

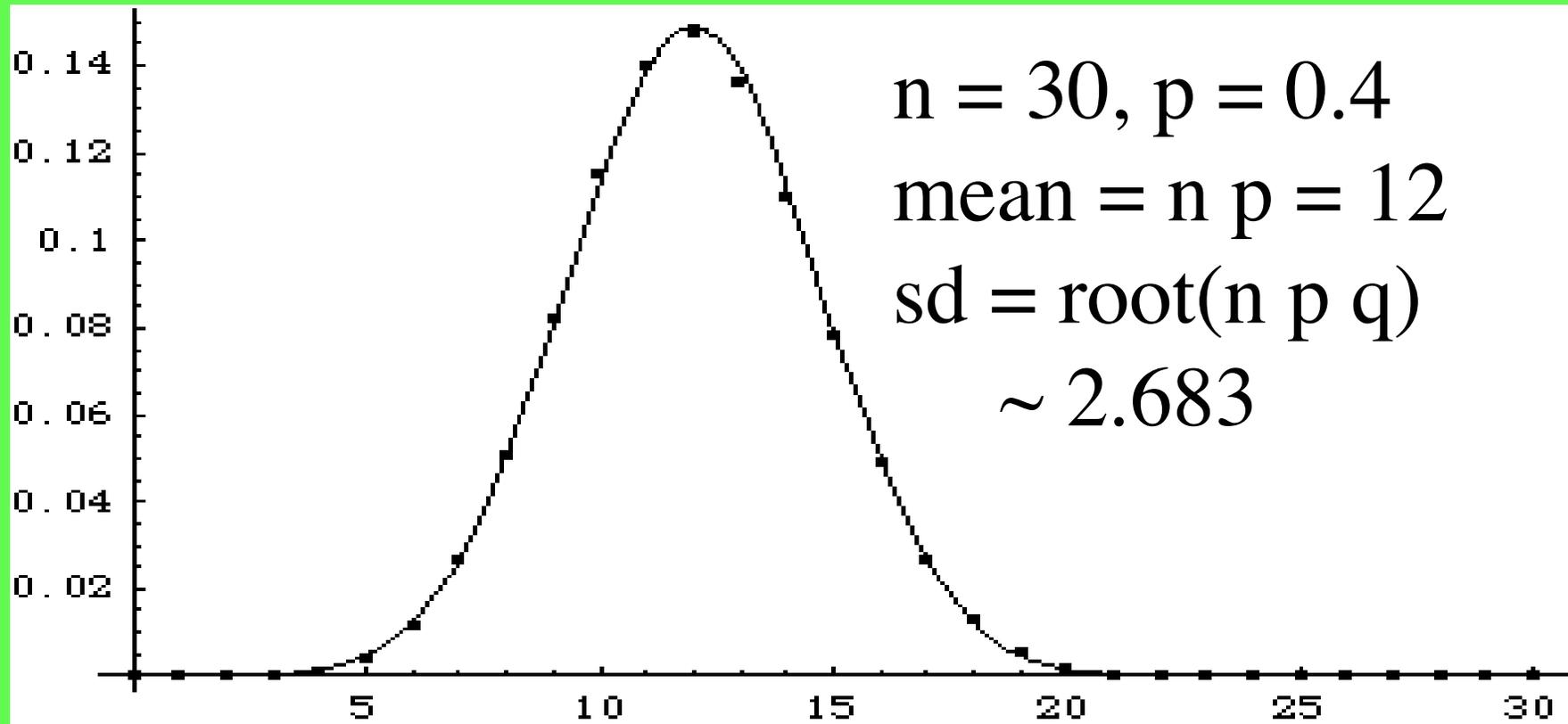
$$p(3) = ((104!)/(3! 101!)) (1/52)^3 (51/52)^{49} \sim 0.182205$$

Agrees with Poisson approximation of binomial!

# Normal Approx of Binomial



# Normal Approx of Binomial



# Normal Approx of Binomial

