

# Lecture 8 Review

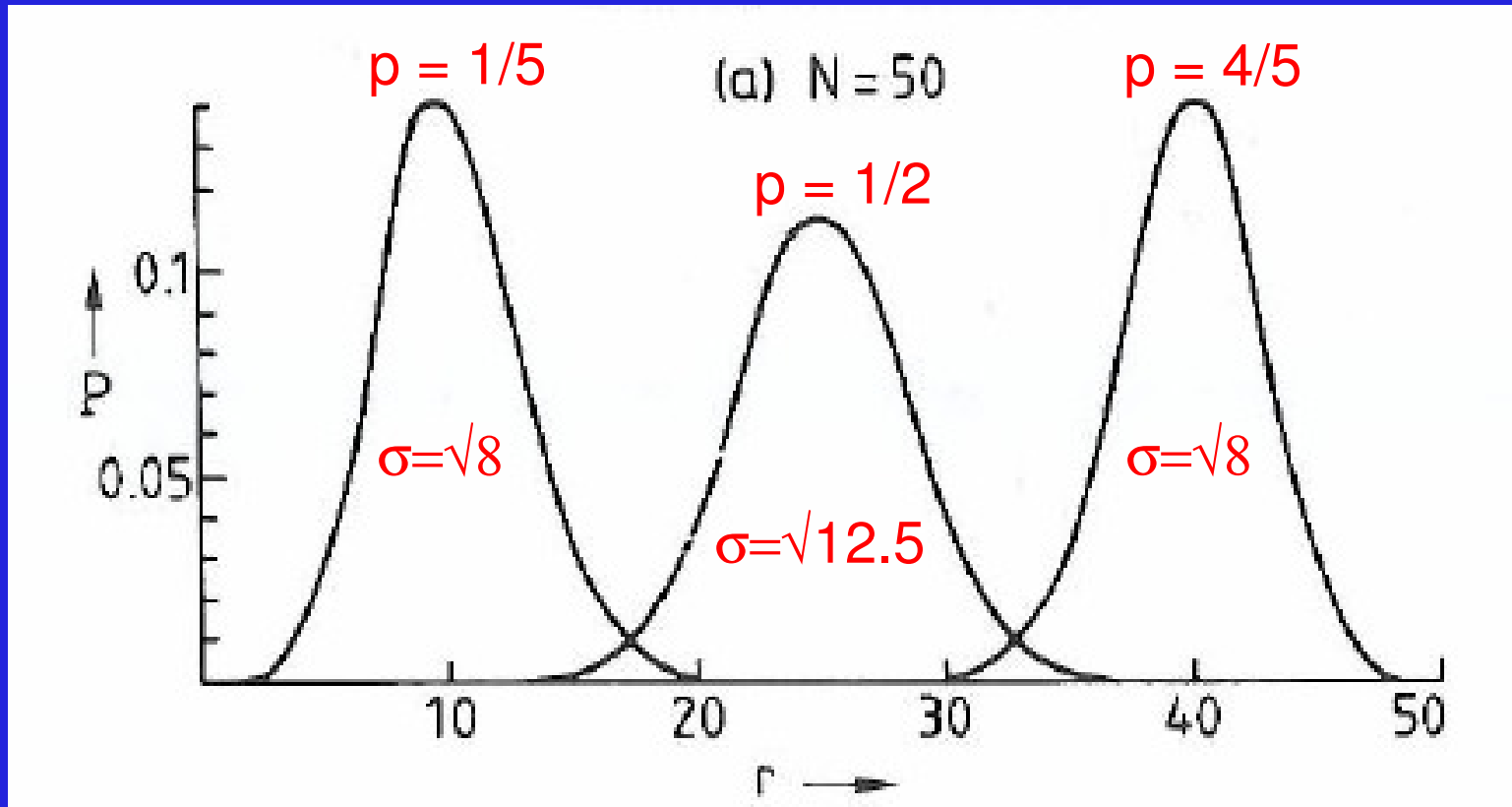
## Binomial probability distribution

$$P_B(x; N, p) = \binom{N}{x} p^x q^{N-x} = \frac{N!}{x!(N-x)!} p^x (1-p)^{N-x}$$

$$\mu = Np \quad \text{mean value}$$

$$\sigma^2 = Npq \quad \text{“variance.”} \quad \sigma \text{ measures width of } P_B.$$

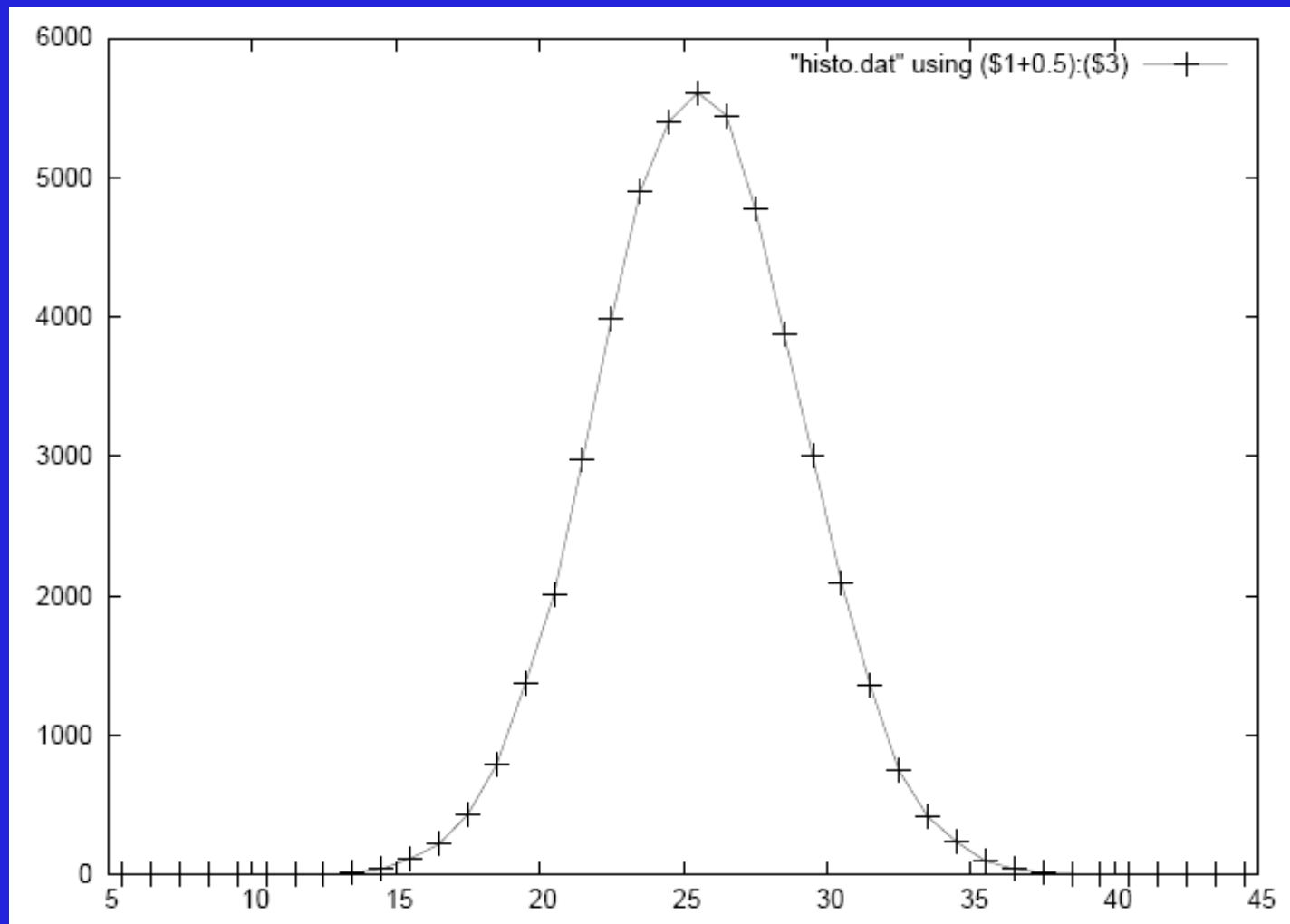
# Binomial Distribution Examples



```
gsl-randist 123 1000 binomial 0.5 50 | gsl-histogram 5 45 40
```

“pipe” symbol: directs output of one command to input of another.

# Binomial Distribution Histogram via gnuplot



```
gnuplot> plot "histo.dat" using ($1+0.5):($3) w lp
```

# Poisson Probability Distribution

- Interesting case when  $N \rightarrow \infty$ ,  $p \rightarrow 0$ , but  $N \cdot p = \text{finite}$ .

$P_B(x; N, p)$  describes prob of observing  $x$  events per unit time out of  $N$  possible, each of which has prob  $p$  of occurring.

$$\lim_{p \rightarrow 0} P_B(x; N, p) = P_P(x; \mu) \equiv \frac{\mu^x}{x!} e^{-\mu}$$

$\mu$  = mean number of events per unit time

$\sigma^2 = \mu$  ( $\sigma$  still characterizes width of probability distribution curve)

**Q:** Plot poisson distribution w/  $\mu = 3.5$ , 10,000 points

Use piping, redirection and `gnuplot`. NO cut-and-paste !!!

# Gaussian Probability Density Distribution

- Another interesting case when  $N \rightarrow \infty$ ,  $p \neq 0$ , but  $N \cdot p \gg 1$   
Same case is obtained from  $P_P(x; \mu)$  when  $\mu \gg 1$

We obtain “gaussian” probability density distribution  $P_G$

$$P_G(x; \sigma, \mu) = \frac{1}{\sqrt{2\pi} \sigma} \exp \left[ -\frac{1}{2} \left( \frac{x-\mu}{\sigma} \right)^2 \right]$$

$P_G$  is a continuous function.  $P_G dx$  is a probability (i.e., a pure number)

$P_G dx$  is a probability of observing  $x$  between  $x$  and  $x + dx$

$\mu$  = mean value of  $x$  in parent distribution

$\sigma^2$  characterizes variation about  $\mu$  of parent distribution.

**Q:** Plot gaussian distribution w/  $\mu = 10.0$ ,  $\sigma = 3.0$ , 10,000 points

Use piping, redirection and `gnuplot`. NO cut-and-paste !!!

# Probability Numbers from Probability Distributions

- Often need to pick a random number from an arbitrary probability distribution  $P(x)$ .

How to do this from uniform probability distribution  $p(r)$  btwn  $[0,1)$ ?

$$p(r) = \begin{cases} 1 & \text{for } 0 \leq r < 1 \\ 0 & \text{otherwise} \end{cases}$$

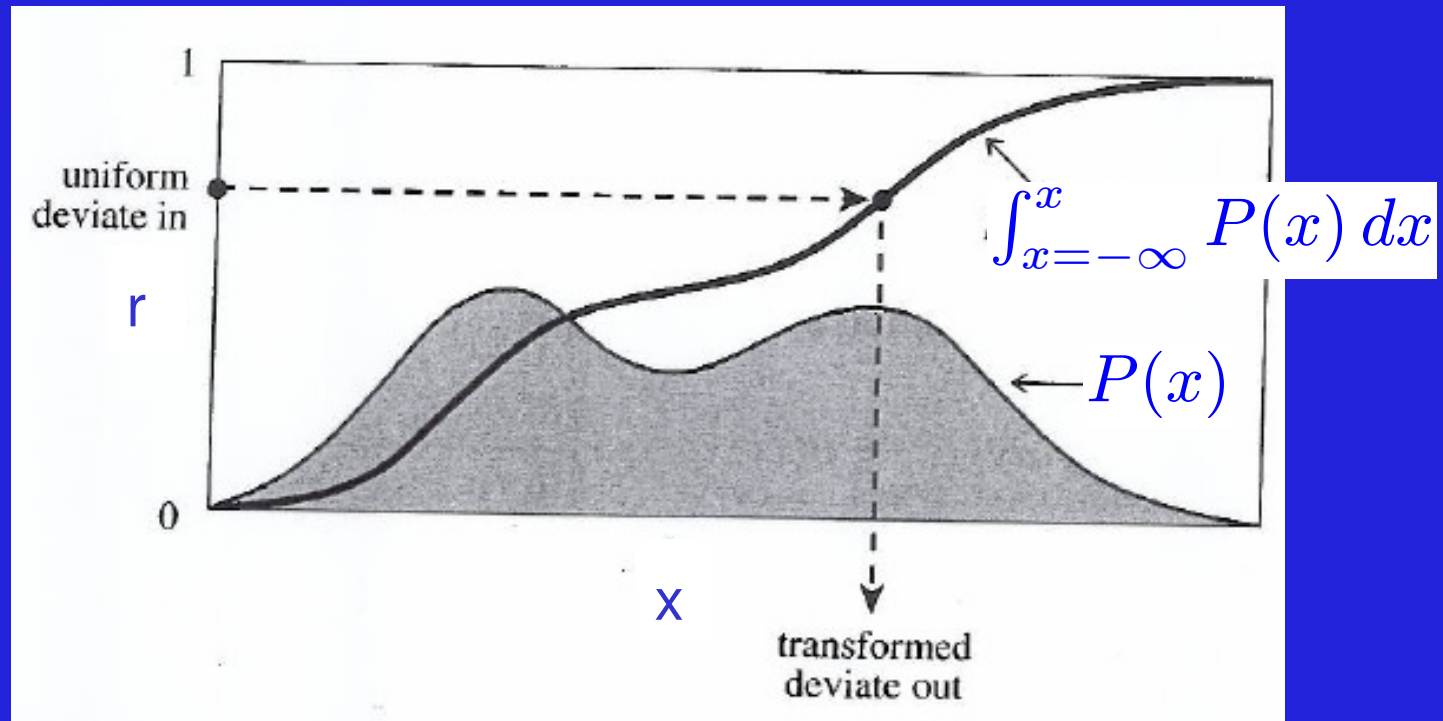
Conservation of probability:  $|p(r)\Delta r| = |P(x)\Delta x|$

$$\int_{r=-\infty}^r p(r) dr = \int_{x=-\infty}^x P(x) dx$$

$$\int_{r=0}^r p(r) dr = \int_{x=-\infty}^x P(x) dx$$

$$r = \int_{x=-\infty}^x P(x) dx$$

# Probability Numbers from Probability Distributions (2)



# Fitting Data (first peek)

Need to fit curves to our data (i.e., fit a curve to a histogram)

Do this by minimizing the quantity  $\chi^2$

$$\chi^2 = \sum_{i=1}^N \left( \frac{y_i - y(x_i)}{\sigma_i} \right)^2$$

$y_i$  is the  $y$  value from the curve.

$y(x_i)$  is the  $y$  value from the data at a particular  $x_i$ .

$\sigma_i$  is the error on the  $y(x_i)$ . (Use  $\text{sqrt}(\text{nmbcr of bin counts})$  in a histogram.)

Example:

```
gnuplot> f1(x) = a1* sin(b1*x); a1 = 1.0; b1 = 0.3;
```

```
gnuplot> fit f1(x) "trash.dat" using ($1):($3) via a1 b1
```



# Summary

Poisson and Gaussian probability distributions.

Pick random numbers from arbitrary probability distribution.

First peek at fitting data.

**Don't suffer in silence. Scream for help!!!**

