

Poisson:  $p(k) = \frac{\mu^k}{k!} e^{-\mu}$

small  $\mu$ ,

$$p(k) \sim \frac{1}{k!}$$

large  $\mu$ ,

$$p(k) \sim \mu^k$$

initially

$$\frac{1}{k!}$$

finally

Gaussian

$$P_G \propto e^{-x^2/2\sigma^2}$$

Cauchy

$$P_c \propto \frac{1}{x^2 + 1}$$

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Previously :

- Gaussian  $\leftrightarrow$  Cauchy.

- Generating random #s.

$$p(x) = \text{pdf}.$$

$$P(x) = \int_{x_{\min}}^x dx' p(x')$$

$$= \text{c. d. f.}$$

$$x_{\text{sim}} = p^{-1}(r) \quad \text{with } r \in [0,1] \text{ uniform.}$$

has pdf  $p(x)$

Binning and cont. dist's.

$$\int_{x_{\min}}^{x_{\max}}$$

$$\Delta x$$

$$f(x) \approx 1 =$$

$$\sum_n$$

$$f_n$$

$$\Delta x$$

Discrete binning

$$\sum_n f_n^{\text{discrete}} = 1$$

code: - generated rand. #

- converted using  $P^{-1}$

- binned

- plotted  $p(x)$

What if we don't know  $p(x)$ ?

If you have normally distributed data, how do you determine  $\sigma$ ?

Many ways to fit function  
to data. A common method  
is by minimizing error

$$e = \left( \sum_n |sim_n - pred_n|^p \right)^{1/p}$$

$p=2$  is "least squares".

If  $x = r^2$ ,  $p(x) = \frac{1}{2\sqrt{x}}$

Take data, and fit to

$$p(x) = ax^n$$

Expect  $n = -\frac{1}{2}$

$$p(x) = ax^n$$

$$\log_b(p(x)) = \log_b(ax^n)$$

$\log y$

$$= n \log_b(x) + \log_b(a)$$

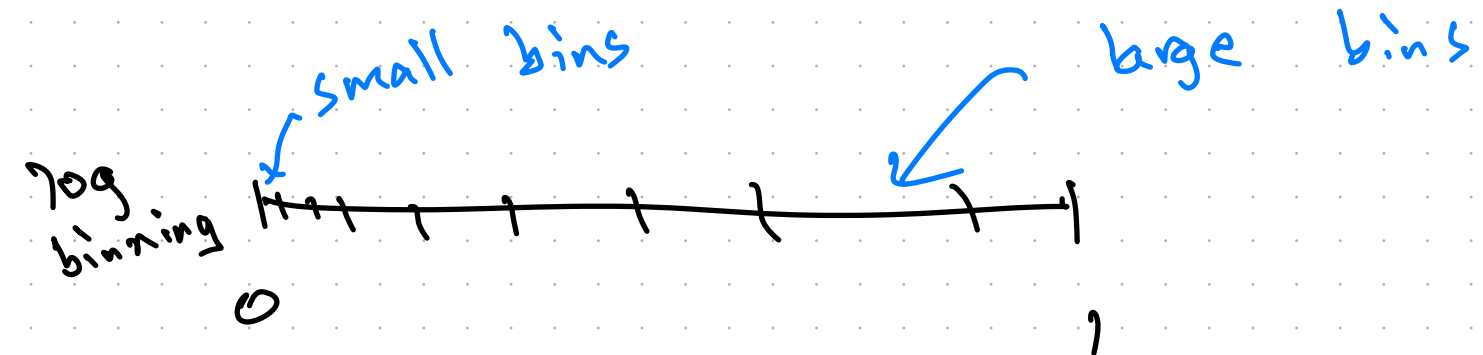
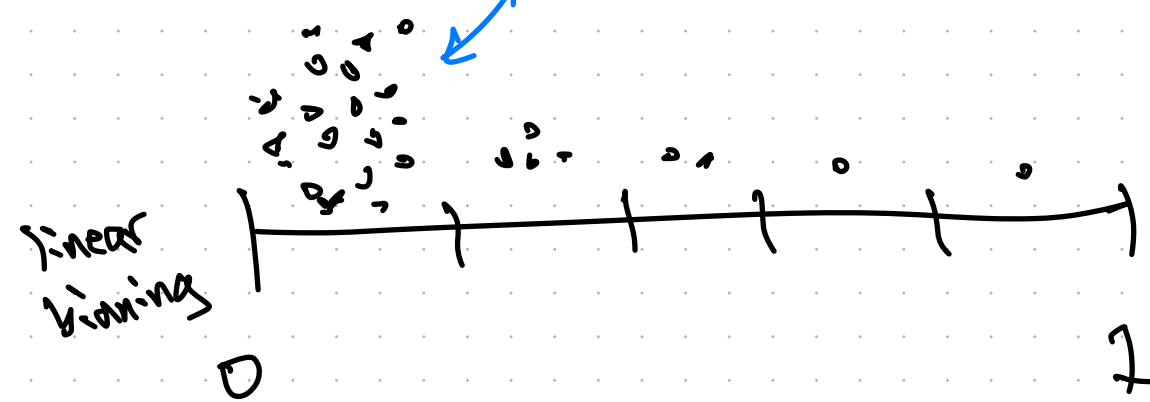
$\log x$

↑  
depends  
on  $b, a$



Our problem

almost all data.



Linear binning:

$$A_n^{\text{discrete}} = \int_{x_n - \frac{\Delta x}{2}}^{x_n + \frac{\Delta x}{2}} dx \, p(x)$$

$$\lim_{\Delta x \rightarrow 0} \int_{x_n - \frac{\Delta x}{2}}^{x_n + \frac{\Delta x}{2}} dx \, p(x) = \Delta x \cdot p(x_n) + O(\Delta x^2)$$

Log binning.  $y = \log x$

$$p_n^{\text{discrete}} = \int_{e^{y_n - \frac{\Delta y}{2}}}^{e^{y_n + \frac{\Delta y}{2}}} dx \, p(x)$$

$$\lim_{\Delta y \rightarrow 0} \int_{e^{y_n - \Delta y/2}}^{e^{y_n + \Delta y/2}} dx \, p(x)$$

$$\lim_{\Delta y \rightarrow 0} \int_{x_n e^{-\Delta y/2}}^{x_n e^{+\Delta y/2}} dx \, p(x)$$

$$= x_n \cdot p(x_n) \cdot \Delta y$$

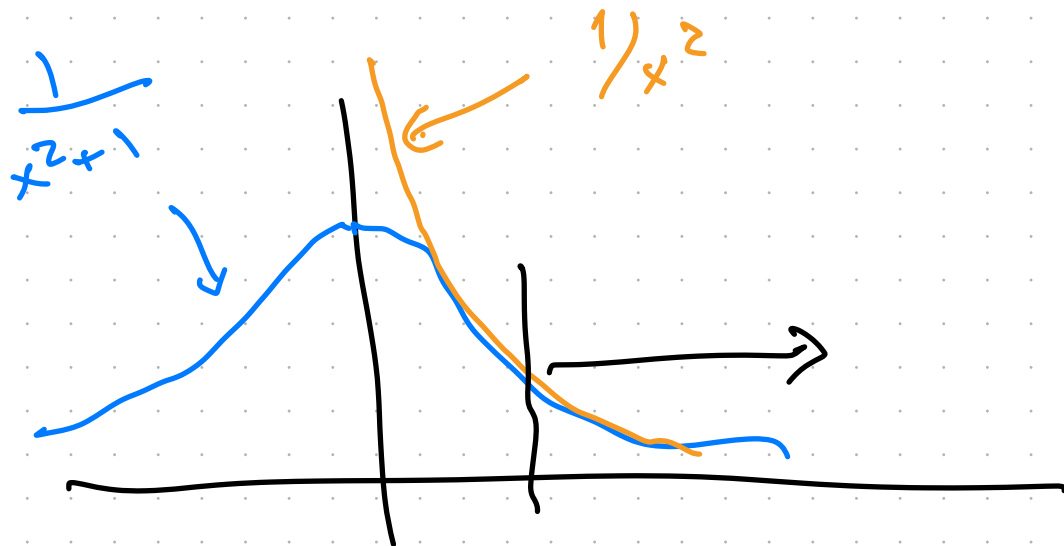


comes from log binning.

For log binned data, if

$$p(x) = \frac{1}{\sqrt{x}}, \quad \text{then}$$

$p_n^{\text{discrete}} \propto \sqrt{x}$  for log binned data



$$x = \frac{1}{\lambda}$$

$$\lambda = \frac{1}{x}$$

$$f(x) = \frac{1}{x}$$

$$p(x) = \frac{1}{x^2}$$

$$x p(x) = \frac{1}{x}$$

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def fcn (x, p1 p2 p3 ...):  
    return p1 · xn1 + p2 xn2
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