

Welcome to S2.

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Statistics

works  
with  
data

Probability

forward  
modelling.  
(uncertain)

Inference.

inference  
problem.

$\theta$  are params of our model

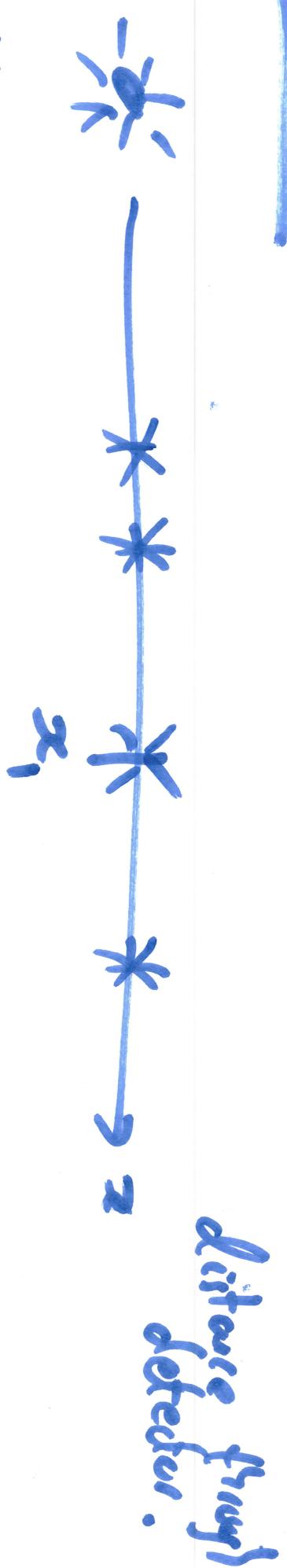
$x$  (random) data seen in experiment.

(believed)  $P(x | \theta)$  prob. of data.

(forward modelling, probabilistic, depends on  $\theta$ ).

aff. notation  $L(\theta)$  (likelihood function

example :



data flesh locations  $\{x_1, x_2, \dots, x_n\}$

$$x_i \sim \text{Exp}(\lambda)$$

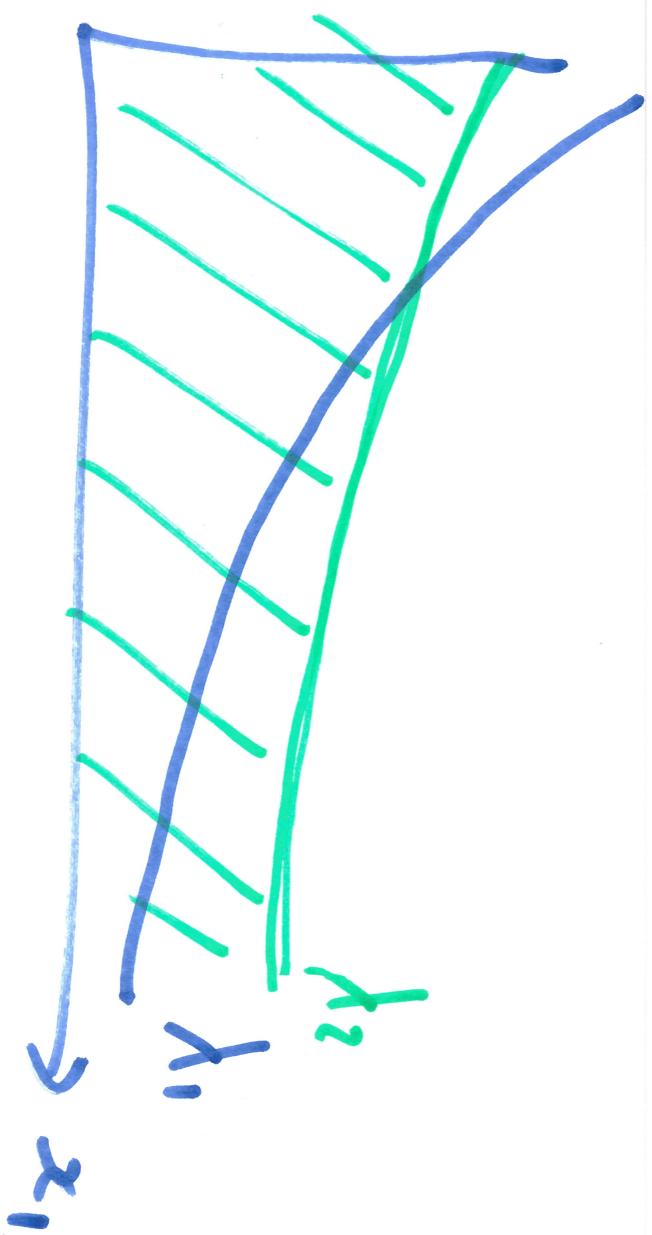
(induced)  $P(x_i | \lambda) = \left\{ \begin{array}{ll} \frac{\exp(-x_i)}{\lambda} & \text{if } x > 0 \\ c & \text{else.} \end{array} \right.$

digital image  
detected.

$$P(x_i | \lambda) = \prod_{i \in I} L(x_i | \lambda)$$

$$P(x_i | \lambda)$$

$$\lambda_2 > \lambda_1$$

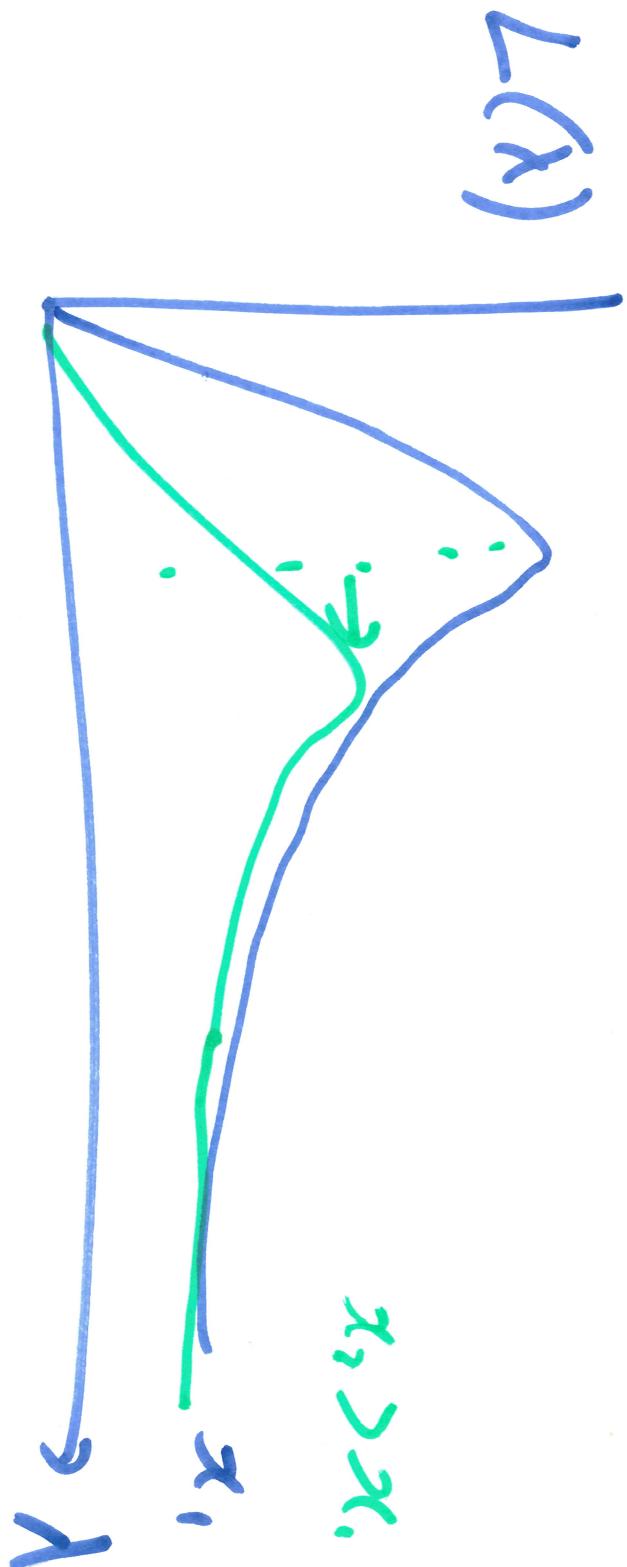


$$\text{normalized w.r.t. } x$$

$$\int_{-\infty}^{\infty} P(x_i | \lambda) dx_i = 1$$

$$P(x_i | \lambda) = \frac{\exp(-x_i/\lambda)}{\lambda}$$

is not normalized w.r.t.  $\lambda$   $\int_0^\infty L(\lambda) d\lambda \neq 1$ .



different possible outcomes

$$P(\{x\} \mid \lambda)$$

$$L(\lambda)$$

$$\frac{*}{f(\{x\} \mid \lambda)}.$$

My notation  
for likelihood.

Example problem

(use on  $\mathcal{M}_m$ ).

estimate # density of stas.

Survey patch of area  $A$ .

Count (all) stas since certain likelihood in  
(affluent) mag.

data ? the number of stas.

↑  
(discrete)

Poisson random variable. assuming independence/uniform density/etc.

$$\mathcal{L}(\sigma | S) = \frac{(\lambda s)^n e^{\lambda s} (-\lambda s)}{n!}$$

num  
density

$$[s] = \deg^{-1}$$

As (dimensionless) expected number.

Use this to find  $s$  in a Bayesian way.  
infer