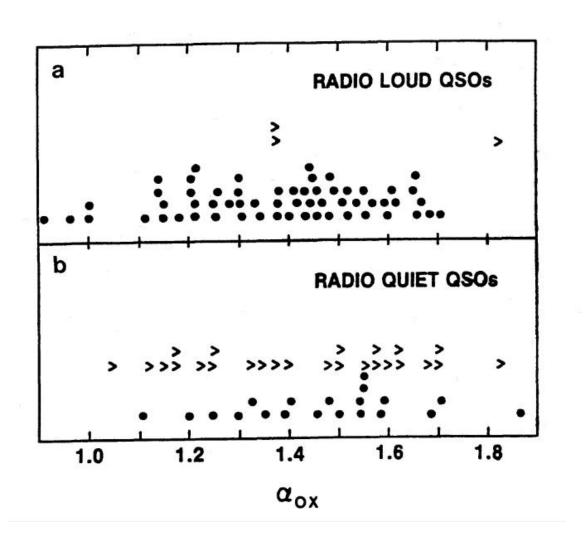
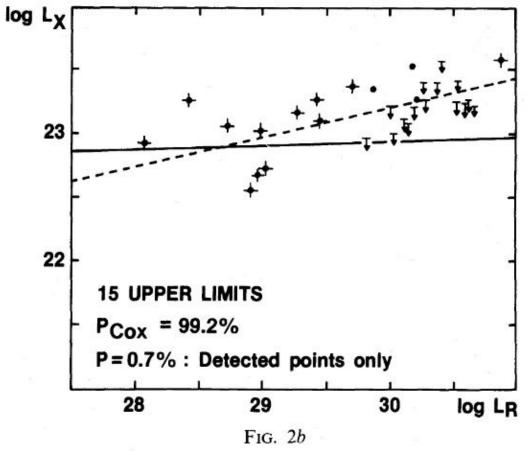
天文探测中上限数据的处理

1984, ApJ, 293, 192

1986, ApJ, 306, 490

问题





Survival analysis

● 天文: censored Data (upper/lower limit)

● 医学: lifetime Data

https://scikit-survival.readthedocs.io

Let T denote a continuous non-negative random variable corresponding to a patient's survival time. The survival function S(t) returns the probability of survival beyond time t and is defined as

$$S(t) = P(T > t).$$

KAPLAN-MEIER PRODUCT-LIMIT ESTIMATOR

$$F(t) = P(T \le t)$$
 .
$$= P(T > t) + P(T = t)$$
 分步积分
$$= 1 - F(t) + P(T = t)$$
 .
$$= \int_0^\infty x dF(x) ,$$

$$= \int_0^\infty S(x) dx .$$

x是人的生存期或者调查时间(调查的时候人还没死)

$$x_{(1)} < x_{(2)} < \dots < x_{(n)}.$$
 Set $x_{(0)} \equiv 0$. For $i = 0, 1, 2, \dots, n-1$, let
$$P_i = P[T \ge x_{(i+1)} | T \ge x_{(i)}]$$

$$S(x_{(j)}) = P[T \ge x_{(j)}],$$

$$= \prod_{i=0}^{j-1} P[T \ge x_{(i+1)} | T \ge x_{(i)}],$$

$$= \prod_{i=0}^{j-1} P_i.$$

The P_i values are estimated as follows. For $i \ge 1$, if $x_{(i)}$ is not a censored value, there are n - i + 1 "true" values $\{x_{(i)}\}$ at least as large as $x_{(i)}$ of which only one $(x_{(i)}$ itself) is not at least as large as $x_{(i+1)}$. In this case, estimate P_i by

$$\hat{P}_i = 1 - 1/(n - i + 1) . {5}$$

If, on the other hand, $x_{(i)}$ is a censored value, it is known that all the true values in the set $\{x_{(j)}\}$ which are at least as large as $x_{(i)}$ are also at least as large as $x_{(i+1)}$. Hence, here estimate P_i by

$$\hat{P}_i = 1 \ . \tag{6}$$

最大似然法:需要假设分布函数f(y|x)

A likelihood function describing a given data set can be defined using the above formulations. Consider a detected point falling in a bin $(z_i, z_i + \Delta z)$. The probability that this occurs is determined by the probability density and is

$$P_D(z_i) \approx f(z_i)\Delta z$$
 (10)

If an object is right censored at z_i , so that the true location of the point is somewhere between z_i and ∞ , the contribution from this point can be written in terms of the survival function

$$P_C(z_j) \approx \int_{z_j}^{\infty} f(t)dt = S(z_j) . \tag{11}$$

If there are m detected observations, and n censored observations, the likelihood function is expressed by

$$L = \prod_{D}^{m} f(z_i) \cdot \prod_{C}^{n} S(z_j) (\Delta z)^{m} ,$$

where \prod_{D}^{m} denotes the product over the *m* detected points, and \prod_{C}^{n} denotes the product over the *n* censored points. Since $(\Delta z)^{m}$ does not contribute to the maximum, the likelihood can be rescaled to be

$$L = \prod_{D}^{m} f(z_i) \prod_{C}^{n} S(z_j) . \tag{12}$$

Taking the logarithm, we get the log likelihood function

$$l = \sum_{D}^{m} \log f(z_i) + \sum_{C}^{n} \log S(z_j) . \tag{13}$$