

失效概率函数相关推导

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①: AK-MCS方法的问题转化

$$\int y^{\alpha_k} f_{Y|\theta_j}(y) dy = M_Y^{\alpha_k}$$

$$\text{mini} : - \int f_{Y|\theta_j}(y) \cdot - \sum_{k=0}^m \lambda_k y^{\alpha_k} dy$$

$$= \sum_{k=0}^m \lambda_k \int f_{Y|\theta_j}(y) y^{\alpha_k} dy$$

$$= + \lambda_0 \int f_{Y|\theta_j}(y) y^0 dy + \sum_{k=1}^m \lambda_k \cdot M_Y^{\alpha_k}$$

$$\xrightarrow{\text{代入 } \alpha_0, \lambda_0 \text{ 则}} \text{mini} = \underbrace{\int f_{Y|\theta_j}(y) dy}_{M_0} \cdot \ln \left[\int_Y \exp \left(- \sum_{k=1}^m \lambda_k y^{\alpha_k} \right) dy \right] + \sum_{k=1}^m \lambda_k M_Y^{\alpha_k}$$

M_0 零阶矩为1

$$\rightarrow \text{mini} \ln \left[\int_Y \exp \left(- \sum_{k=1}^m \lambda_k y^{\alpha_k} \right) dy \right] + \sum_{k=1}^m \lambda_k M_Y^{\alpha_k}$$

②. 乘法降维:

首先, 由 $y(x) = \ln |g(x)|$, 代入近似有:

$$y(x) \approx \sum_{i=1}^n \ln |g(x_i, \mu_{\sim x_i})| - (n-1) \ln |g(x_0)|$$

两边取指数:

$$\exp(y(x)) = g(x) = \frac{\prod_{i=1}^n g(x_i, \mu_{\sim x_i})}{g(x_0)^{(n-1)}}$$

$$\xrightarrow{\text{整理}} g(x) = g(x_0)^{1-n} \cdot \prod_{i=1}^n g(x_i, \mu_{\sim x_i})$$

$$\xrightarrow{\text{近似}} g(x) = g(x_0)^{1-n} \cdot \prod_{i=1}^n g(x_i, \mu_{\sim x_i})$$

由定义Y分数矩为:

$$M_Y^{\alpha_R} = \int y^{\alpha_R} f_{Y|\theta_j}(y) dy$$

\Rightarrow 其中 $Y = g(X)$, 则:

$$M_{Y|\theta_j}^{\alpha_R} = \int_{x|\theta_j} g(x)^{\alpha_R} f_{x|\theta_j}(x) dx, \text{代入:}$$

$$= \int_x g(x_0)^{\alpha_R(1-n)} \cdot \prod_{i=1}^n (g(x_i, \mu_{\sim x_i}))^{\alpha_R} f_{x|\theta_j}(x) dx$$

$$\xrightarrow{x_0 = \mu_x} \approx g(\mu_x)^{\alpha_R(1-n)} \cdot \prod_{i=1}^n \int_{x_i} f_{x|\theta_j}(x) \cdot (g(x_i, \mu_{\sim x_i}))^{\alpha_R} dx$$

即变成了 n个单变量积分连乘

而积分可使用高斯积分公式近似求解