

# 参数估计

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## 1 判断观测数据中是否有信号

求  $\mathbb{P}(A_m|A_g)$

$$G_t(\omega) = \begin{cases} N_t(\omega) & \omega \in A_0 \\ N_t(\omega) + M_t(\omega) & \omega \in A_m \end{cases}$$

$$\mathbb{P}(A_m|A_g) = \frac{\mathbb{P}(A_g|A_m)\mathbb{P}(A_m)}{\mathbb{P}(A_g)}$$

$$\begin{aligned} \mathbb{P}(A_g) &= \mathbb{P}(A_g|A_0)\mathbb{P}(A_0) + \mathbb{P}(A_g|A_m)\mathbb{P}(A_m) \\ &= \mathbb{P}(A_g|A_0)\mathbb{P}(A_0) + \mathbb{P}(A_m) \frac{\mathbb{P}(A_g \cap A_m)}{\mathbb{P}(A_m)} \\ &= \mathbb{P}(A_g|A_0)\mathbb{P}(A_0) + \mathbb{P}(A_m) \frac{\int d\boldsymbol{\mu} p(A_g \cap A_\mu)}{\mathbb{P}(A_m)} \\ &= \mathbb{P}(A_g|A_0)\mathbb{P}(A_0) + \mathbb{P}(A_m) \int d\boldsymbol{\mu} \frac{p(A_g \cap A_\mu)}{p(A_\mu)} \frac{p(A_\mu)}{\mathbb{P}(A_m)} \\ &= \mathbb{P}(A_g|A_0)\mathbb{P}(A_0) + \mathbb{P}(A_m) \int d\boldsymbol{\mu} \mathbb{P}(A_g|A_\mu)p(\boldsymbol{\mu}) \end{aligned}$$

$$\mathbb{P}(A_m|A_g) = \frac{\Lambda}{\Lambda + \mathbb{P}(A_0)/\mathbb{P}(A_m)}$$

$$\Lambda := \int d\boldsymbol{\mu} \lambda(\boldsymbol{\mu})$$

$$\lambda(\boldsymbol{\mu}) := p(\boldsymbol{\mu}) \frac{\mathbb{P}(A_g | A_{\boldsymbol{\mu}})}{\mathbb{P}(A_g | A_0)}$$

**定理 1.**  $\mathbb{P}(A_g | A_{\boldsymbol{\mu}}) = \mathbb{P}(A_{g-m(\boldsymbol{\mu})} | A_0) \Leftrightarrow \{N = n\}$  独立于  $A_0, A_{\boldsymbol{\mu}}$ .

证明.  $G|_{A_{\boldsymbol{\mu}}} = N|_{A_{\boldsymbol{\mu}}} + m(\boldsymbol{\mu}) \Rightarrow \mathbb{P}(A_g | A_{\boldsymbol{\mu}}) = \frac{p(\{G=g\} \cap A_{\boldsymbol{\mu}})}{p(A_{\boldsymbol{\mu}})} = \frac{p(\{N=g-m(\boldsymbol{\mu})\} \cap A_{\boldsymbol{\mu}})}{p(A_{\boldsymbol{\mu}})}$ .

$G|_{A_0} = N|_{A_0} \Rightarrow \mathbb{P}(A_{g-m(\boldsymbol{\mu})} | A_0) = \frac{\mathbb{P}(\{G=g-m(\boldsymbol{\mu})\} \cap A_0)}{\mathbb{P}(A_0)} = \frac{\mathbb{P}(\{N=g-m(\boldsymbol{\mu})\} \cap A_0)}{\mathbb{P}(A_0)}$ .

$\frac{\mathbb{P}(\{N=n\} \cap A_0)}{\mathbb{P}(A_0)} = \frac{p(\{N=n\} \cap A_{\boldsymbol{\mu}})}{p(A_{\boldsymbol{\mu}})} := k$

$\Rightarrow \mathbb{P}(\{N = n\} \cap A_0) + \int d\boldsymbol{\mu} p(\{N = n\} \cap A_{\boldsymbol{\mu}}) = k\mathbb{P}(A_0) + \int d\boldsymbol{\mu} kp(A_{\boldsymbol{\mu}})$

$\Rightarrow \mathbb{P}(\{N = n\}) = k(\mathbb{P}(A_0) + \int d\boldsymbol{\mu} p(A_{\boldsymbol{\mu}})) = k$ .

依充分性的证明方法, 必要性是易证的. □

$$\lambda(\boldsymbol{\mu}) := p(\boldsymbol{\mu}) \frac{\mathbb{P}(A_{g-m(\boldsymbol{\mu})} | A_0)}{\mathbb{P}(A_g | A_0)}$$

$$S_n(f) = 2\mathcal{F}(C_n(\tau))$$

$$\mathbb{P}(A_{\{g_i\}} | A_0) = \frac{\exp[-\frac{1}{2} \sum C_{jk}^{-1} g_j g_k]}{[(2\pi)^N \det(C_{jk})]^{1/2}}$$

$$\begin{aligned} & \int_{-\infty}^{\infty} dt_j e^{2\pi i f t_j} \int_{-\infty}^{\infty} dt_l C_n(t_j - t_l) C^{-1}(t_l, t_k) \\ &= \int_{-\infty}^{\infty} dt_j \int_{-\infty}^{\infty} dt_l C_n(t_j - t_l) C^{-1}(t_l, t_k) e^{2\pi i f t_j} \\ &= \int_{-\infty}^{\infty} dt_l \int_{-\infty}^{\infty} dt_j C_n(t_j - t_l) C^{-1}(t_l, t_k) e^{2\pi i f t_j} \\ &= \int_{-\infty}^{\infty} dt_l \int_{-\infty}^{\infty} d\tau C_n(\tau) C^{-1}(t_l, t_k) e^{2\pi i f t_l} e^{2\pi i f \tau} \quad (t_j - t_l = \tau) \\ &= \int_{-\infty}^{\infty} dt_l C^{-1}(t_l, t_k) e^{2\pi i f t_l} \int_{-\infty}^{\infty} d\tau C_n(\tau) e^{2\pi i f \tau} \end{aligned}$$

**定理 2.**  $\int_{-\infty}^{\infty} dt p(t)q(t) = \int_{-\infty}^{\infty} df \mathcal{F}(p)(f)\mathcal{F}(q)(f)^* \Leftrightarrow q \in \mathbb{R}.$

证明.  $\int_{-\infty}^{\infty} dt p(t)q(t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} dt df p(t)\mathcal{F}(q)(f)e^{-2\pi i f t}.$   
 $\int_{-\infty}^{\infty} df \mathcal{F}(p)(f)\mathcal{F}(q)(f)^* = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} df dt p(t)\mathcal{F}(q)(f)^* e^{2\pi i f t},$   
 $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} df dt p(t)\mathcal{F}(q)(f)^* e^{2\pi i f t} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} df dt p(t)\mathcal{F}(q)(-f)^* e^{-2\pi i f t}.$   
取  $\int_{-\infty}^{\infty} dt p(t)e^{-2\pi i f t} = \delta(f - f_0)$ , 即  $p(t) = e^{2\pi i f_0 t}$ ,  
得  $\mathcal{F}(q)(f) = \mathcal{F}(q)(-f)^* \Rightarrow q \in \mathbb{R}.$   
依充分性的证明方法, 必要性是易证的. □

$$\mathbb{P}(A_m|A_g) = \frac{\Lambda}{\Lambda + \mathbb{P}(A_0)/\mathbb{P}(A_m)}$$

$$\Lambda := \int d\boldsymbol{\mu} \lambda(\boldsymbol{\mu})$$

$$\lambda(\boldsymbol{\mu}) := p(\boldsymbol{\mu}) \exp[2 \langle g(t), m_{\boldsymbol{\mu}}(t) \rangle - \langle m_{\boldsymbol{\mu}}(t), m_{\boldsymbol{\mu}}(t) \rangle]$$

$$\langle p(t), q(t) \rangle = \int_{-\infty}^{\infty} df \frac{\tilde{p}(f)\tilde{q}(f)^*}{S_n(f)}$$

## 2 参数估计 (MLE)

求  $\hat{\boldsymbol{\mu}}$  和  $V(P)$

$$p(A_{\boldsymbol{\mu}}|A_g) = \frac{\lambda(\boldsymbol{\mu})}{\Lambda + \mathbb{P}(A_0)/\mathbb{P}(A_m)}$$

$$p(A_{\boldsymbol{\mu}}|A_g \cap A_m) = \frac{\lambda(\boldsymbol{\mu})}{\Lambda}$$

$$\frac{\partial \ln p(\boldsymbol{\mu})}{\partial \boldsymbol{\mu}}|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}} + 2 \left\langle \frac{\partial m_{\boldsymbol{\mu}}}{\partial \boldsymbol{\mu}}|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}}(t), g(t) - m_{\boldsymbol{\mu}}|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}}(t) \right\rangle = 0$$

$$P = \int_{\boldsymbol{\mu} \in V(P)} d\boldsymbol{\mu} p(A_{\boldsymbol{\mu}}|A_g)$$

$$V(P) = \{\boldsymbol{\mu} \mid p(A_{\boldsymbol{\mu}}|A_g) \geq \kappa^2\}$$

### 3 灵敏度 (准确度)

#### 3.1 精确结果

求  $p(A_{\hat{\mu}}|A_{\mu})$

$$A_{\hat{\mu}} = \bigcup_{g \Rightarrow \hat{\mu}} A_g$$

$$\nu_i = 2 \left\langle N_t, \frac{\partial m_{\mu}}{\partial \mu_i} \Big|_{\mu=\hat{\mu}}(t) \right\rangle \quad (A_{\hat{\mu}} \rightarrow \mathbb{R})$$

$$\langle N_t, h(t) \rangle = \int_{-\infty}^{\infty} df \frac{\mathcal{F}(N)(f) \mathcal{F}(h)(f)^*}{S_n(f)}$$

$$\mathcal{F}(N)(f) = \int_{-\infty}^{\infty} dt N_t e^{2\pi i f t}$$

**定理 3.** 常数乘任意期望为0的正态随机变量后是期望为0的正态随机变量.

**定理 4.** 任意两个期望为0的正态随机变量的和是期望为0的正态随机变量.

**定理 5.** 任意期望为0的正态随机变量序列, 若其极限存在, 则此极限是期望为0的正态随机变量.

**定理 6.**

$$\begin{aligned} & \int_{-\infty}^{\infty} dt \left( \int_{-\infty}^{\infty} df W(f) e^{-2\pi i f t} \right) \left( \int_{-\infty}^{\infty} df' W'(f') e^{-2\pi i f' t} \right) \\ &= \int_{-\infty}^{\infty} df W(f) \left( \int_{-\infty}^{\infty} df' W'(f') \delta(f + f') \right). \end{aligned}$$

证明.

$$\begin{aligned} & \int_{-\infty}^{\infty} dt \left( \int_{-\infty}^{\infty} df W(f) e^{-2\pi i f t} \right) \left( \int_{-\infty}^{\infty} df' W'(f') e^{-2\pi i f' t} \right) \\ &= \int_{-\infty}^{\infty} dt \mathcal{F}(W)(t) \mathcal{F}(W')(t). \end{aligned}$$

$$\begin{aligned}
& \int_{-\infty}^{\infty} df W(f) \left( \int_{-\infty}^{\infty} df' W'(f') \delta(f + f') \right) \\
&= \int_{-\infty}^{\infty} df W(f) W'(-f). \\
& \int_{-\infty}^{\infty} dt \mathcal{F}(W)(t) \mathcal{F}(W')(t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} dt df W(f) \mathcal{F}(W')(t) e^{-2\pi i f t}. \\
& \int_{-\infty}^{\infty} df W(f) W'(-f) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} df dt W(f) \mathcal{F}(W')(t) e^{2\pi i (-f)t}.
\end{aligned}$$

□

$$\begin{aligned}
& S_n(f) \\
&= 2 \int_{-\infty}^{\infty} d\tau C_n(\tau) e^{2\pi i f \tau} \\
&= 2 \int_{-\infty}^{\infty} d\tau \left[ \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T dt n(t) n(t + \tau) \right] e^{2\pi i f \tau} \\
&= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} \int_{-T}^T d\tau dt n(t) n(t + \tau) e^{2\pi i f \tau}
\end{aligned}$$

$$\begin{aligned}
& |\mathcal{F}(n)(f)|^2 \\
&= \left[ \int_{-\infty}^{\infty} dt n(t) e^{-2\pi i f t} \right] \left[ \int_{-\infty}^{\infty} dt' n(t') e^{2\pi i f t'} \right] \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} dt' dt n(t) n(t') e^{2\pi i f (t' - t)}
\end{aligned}$$

$$\nu_i(\omega) = -2 \left\langle m_{\hat{\mu}(\omega)}(t) - m_{\hat{\mu}}(t), \frac{\partial m_{\boldsymbol{\mu}}}{\partial \mu_i} \Big|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}(t)} \right\rangle - \frac{\partial \ln p(\boldsymbol{\mu})}{\partial \boldsymbol{\mu}} \Big|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}}$$

$$p(A_{\nu}|A_{\hat{\mu}}) = \frac{\exp[-\frac{1}{2} \sum \mathcal{C}_{ij} \nu_i \nu_j]}{[(2\pi)^N \det(\mathcal{C}_{ij}^{-1})]^{1/2}}$$

$$\mathcal{C}_{ij}^{-1} = 2 \left\langle \frac{\partial m_{\boldsymbol{\mu}}}{\partial \mu_i} \Big|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}(t)}, \frac{\partial m_{\boldsymbol{\mu}}}{\partial \mu_j} \Big|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}(t)} \right\rangle$$

$$P = \int_{V(P)} d\tilde{\boldsymbol{\mu}} p(A_{\tilde{\boldsymbol{\mu}}} | A_{\hat{\boldsymbol{\mu}}}) = \int_{\boldsymbol{\nu}(V(P))} d\boldsymbol{\nu} p(A_{\boldsymbol{\nu}} | A_{\hat{\boldsymbol{\mu}}})$$

$$V(P) = \{\tilde{\boldsymbol{\mu}} | p(A_{\boldsymbol{\nu}(\tilde{\boldsymbol{\mu}})} | A_{\hat{\boldsymbol{\mu}}}) \geq \mathcal{K}^2\}$$

### 3.2 近似结果

$$\delta\mu_i(\omega) = - \sum \mathcal{C}_{ij} \left[ \nu_j(\omega) + \frac{\partial \ln p(\boldsymbol{\mu})}{\partial \mu_j} \Big|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}} \right]$$

$$\overline{\left( \sum \mathcal{C}_{ik} \nu_k \right) \left( \sum \mathcal{C}_{jl} \nu_l \right)} = \overline{\sum \mathcal{C}_{ik} \nu_k \nu_l \mathcal{C}_{lj}} = \sum \mathcal{C}_{ik} \overline{\nu_k \nu_l} \mathcal{C}_{lj} = \sum \mathcal{C}_{ik} \mathcal{C}_{kl}^{-1} \mathcal{C}_{lj} = \mathcal{C}_{ij}$$

$$p(A_{\hat{\boldsymbol{\mu}}+\delta\boldsymbol{\mu}} | A_{\hat{\boldsymbol{\mu}}}) = \frac{\exp[-\frac{1}{2} \sum \mathcal{C}_{ij}^{-1} (\delta\mu_i - \overline{\delta\mu_i})(\delta\mu_j - \overline{\delta\mu_j})]}{[(2\pi)^N \det(\mathcal{C}_{ij})]^{1/2}}$$

$$\overline{\delta\mu_i} = - \sum \mathcal{C}_{ij} \frac{\partial \ln p(\boldsymbol{\mu})}{\partial \mu_j} \Big|_{\boldsymbol{\mu}=\hat{\boldsymbol{\mu}}}$$