

Mathematical foundation of the inverse problem approach

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Based on [ASX, Ting Wei(魏婷), Fu-Sheng Yu(于福升)] arXiv: 2211.13753



目录

- 研究动机
- 反问题介绍
- 适定性研究
- 正则化方法
- 数值结果
- 总结与展望





重大科学难题:非微扰物理量难以计算, 是世纪难题

夸克禁闭和新物理……

现有的非微扰方法各有优缺点

格点QCD(LQCD): 第一性原理的计算方法, 但需要 超级计算机且对激发态等物理量计算难度大。

其他方法:如QCD求和规则、Dyson-Schwinger方程 等都有模型依赖性。唯象方法预言能力有限。

发展新的方法"反问题方法"计算非微扰物理量

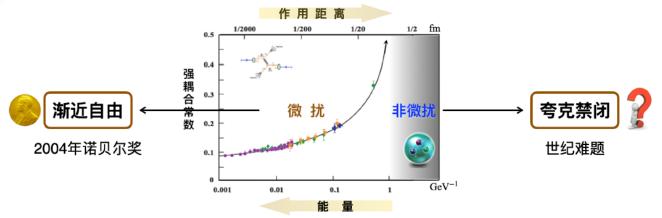
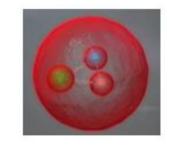
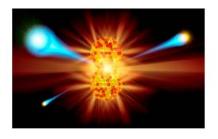


图1: 高能微扰与低能非微扰

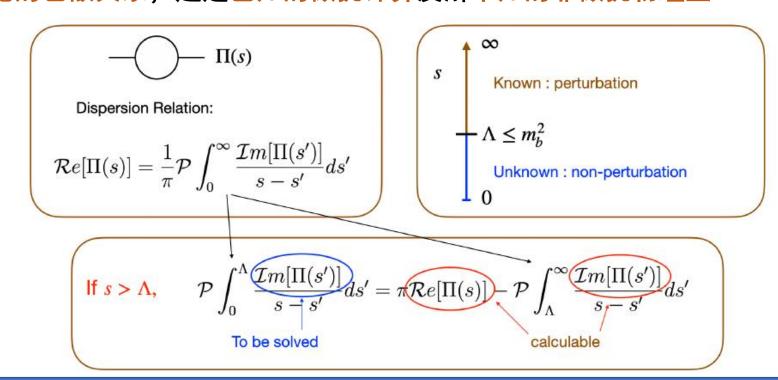




反问题介绍



基于量子场论的色散关系,通过已知的微扰计算反解未知的非微扰物理量:



H.N.Li, H.U, F.R.Xu, **F.S.Yu**, Phys.Lett.B 810(2020)

$$Kf = \int_a^b \frac{f(x)}{y - x} dx = g(y), y \in [c, d], c > b$$

已知积分结果求被积函数需要用到反问题理论。

反问题介绍



反问题理论是成熟的数学分支领域:

与反问题对比的是正问题;

1950年左右发展至今;

在众多重要的前沿科学领域中广泛运用;

线性方程组的求解、地质探勘、缪子成像、CT······

$$\text{If } s > \Lambda, \qquad \mathcal{P} \int_0^{\Lambda} \underbrace{\mathcal{I}m[\Pi(s')]}_{s-s'} ds' = \pi \underbrace{\mathcal{R}e[\Pi(s)]}_{\text{Calculable}} - \mathcal{P} \int_{\Lambda}^{\infty} \underbrace{\mathcal{I}m[\Pi(s')]}_{s-s'} ds'$$

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), y \in [c, d], c > b$$

微扰论





色散关系 反问题方法 非微扰 物理量



色散关系的算子 $K: F(L^2(a,b)) \rightarrow G(L^2(c,d))$

$$\int_{a}^{b} \frac{f(x)}{y-x} dx = g(y), y \in [c,d], c > b$$

正问题: Kf = g

反问题: $f = K^{-1}g$

适定性是研究反问题的基础:

Define: The operator equation (3.1) is called well-posed if the following holds [8]:

1. Existence: For every $g \in G$ there is (at least one) $f \in F$ such that Kf = g;

2. Uniqueness: For every $g \in G$ there is at most one $f \in F$ with Kf = g;

3. Stability: The solution f depends continuously on g; that is, for every sequence $(f_n) \subset F$ with

 $Kf_n \to Kf(n \to \infty)$, it follows that $f_n \to f(n \to \infty)$

若上述有一个条件不满足,就称该问题是不适定问题。

存在唯一性等价于逆算子 K^{-1} : $G \to F$ 存在,稳定性等价于逆算子 K^{-1} 连续(有界)。



反问题的两个简单例子:

- $Formula Kx = 10^{-6}x = y, x \in R, y \in R$ 。其反问题是 $x = K^{-1}y = 10^{6}y$ 。 x = 1, y =
- \triangleright 线性方程组的一般形式为: $A_{n\times n}x_{n\times 1} = b_{n\times 1}$ 。

上述方程组的解存在当且仅当 $r_A = r_{(A,b)}$; 当解存在时,解唯一当且仅当 $r_A = n$;

$$\begin{cases} 2x_1 + 3x_2 = 8\\ 2x_1 + 3.00001x_2 = 8.00002 \end{cases}$$

解为 $x_1 = 1, x_2 = 2;$

当对方程组的右端加上微小的误差时

$$\begin{cases} 2x_1 + 3x_2 = 8\\ 2x_1 + 3.00001x_2 = 8.00003 \end{cases}$$

解却变成 $x_1 = -0.5$, $x_2 = 3$ 。解发生了巨大的变化, 其逆矩阵 A^{-1} 行列式很大。



解的存在性:

R(K) is the analytic functions

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), y \in [c, d], c > b$$

$$Kf = \int_{a}^{b} \frac{f(x)}{y - x} dx = \int_{a}^{b} \frac{1}{y} \frac{f(x)}{1 - \frac{x}{y}} dx$$
$$= \frac{1}{y} \int_{a}^{b} \sum_{k=0}^{\infty} (\frac{x}{y})^{k} f(x) dx = \sum_{k=0}^{\infty} \frac{1}{y^{k+1}} \int_{a}^{b} x^{k} f(x) dx$$

where the last equal sign uses the control convergence theorem:

$$\sum_{k=0}^{\infty} \frac{1}{y^{k+1}} \int_{a}^{b} x^{k} f(x) dx \leq \sum_{k=0}^{\infty} \frac{1}{y^{k+1}} \int_{a}^{b} x^{2k} dx ||f(x)||_{L^{2}(a,b)}$$

$$\leq \sum_{k=0}^{\infty} \frac{1}{y^{k+1}} b^{k} \sqrt{b-a} ||f(x)||_{L^{2}(a,b)}$$

$$\leq \sum_{k=0}^{\infty} \frac{1}{y^{k+1}} (\frac{b}{c})^{k} ||f(x)||_{L^{2}(a,b)} \leq \infty$$

Thus, the R(K) is the analytic functions $\in [c,d]$



解的存在性:



解的唯一性:



Theorem 3.3. Suppose that $f_1(x)$, $f_2(x) \in L^2(a,b)$. If $Kf_1 = Kf_2 = g(y)$, $y \in [c,d]$, then we have $f_1(x) = f_2(x)$, $a. e. x \in [a,b]$.

Proof. Since K is a linear operator, we know that $Kf_1 - Kf_2 = K(f_1 - f_2) = 0$. Therefore, in order to prove $f_1(x) = f_2(x)$, a. e. $x \in [a, b]$, we just need to prove that Kf = 0 implies f(x) = 0, a. e. $x \in [a, b]$.

It is easy to obtain that $Kf = \int_a^b \frac{1}{y-x} f(x) dx = \int_a^b \left(\frac{1}{y} \sum_{k=0}^\infty (\frac{x}{y})^k \right) f(x) dx$. Since $x \in [a,b], y \in [c,d],$ c > b, we know $|\frac{x}{y}| \le |\frac{b}{c}| < 1$, which implies that $\left| \sum_{k=0}^\infty (\frac{x}{y})^k f(x) \right| \le \sum_{k=0}^\infty (\frac{b}{c})^k |f(x)|$ for all $x \in [a,b]$. Combined with $\int_a^b |f(x)| dx < +\infty$ and the control convergence theorem, we have

$$y \int_{a}^{b} \frac{1}{y - x} f(x) dx = \sum_{k=0}^{\infty} \frac{1}{y^{k}} \int_{a}^{b} x^{k} f(x) dx = 0, \quad y \in [c, d].$$
 (3.4)

If $d = +\infty$, by using (3.4), we have

$$\int_{a}^{b} f(x)dx + \frac{1}{y} \int_{a}^{b} x f(x)dx + \dots + \frac{1}{y^{k}} \int_{a}^{b} x^{k} f(x)dx + \dots = 0, \quad y \in (c, +\infty).$$
 (3.5)

Letting $y \to +\infty$ in (3.5), we have $\int_a^b f(x)dx = 0$. Then multiplying y on both sides of (3.5) and letting $y \to +\infty$, we also have $\int_a^b x f(x)dx = 0$. Repeating above process, we can obtain that

$$\int_{a}^{b} x^{k} f(x) dx = 0, \quad k = 0, 1, 2, \cdots.$$
 (3.6)

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), y \in [c, d], c > b$$

By using (3.6), we know that $\int_a^b f(x)Q_n(x)dx = 0$. Combined with the Cauchy inequality, we have

$$\begin{split} \|f\|_{L^{2}(a,b)}^{2} &= \int_{a}^{b} f^{2}(x) dx = \int_{a}^{b} \left(f^{2}(x) - f(x) Q_{n}(x) \right) dx \\ &\leq \int_{a}^{b} |f(x)| \cdot |f(x) - Q_{n}(x)| dx \\ &\leq \left(\int_{a}^{b} f^{2}(x) dx \right)^{\frac{1}{2}} \left(\int_{a}^{b} |f(x) - Q_{n}(x)|^{2} dx \right)^{\frac{1}{2}} \\ &= \|f\|_{L^{2}(a,b)} \|f - Q_{n}\|_{L^{2}(a,b)} \\ &\leq (\epsilon + \epsilon \sqrt{b - a}) \|f\|_{L^{2}(a,b)}, \end{split}$$

which implies that $||f||_{L^2(a,b)} \le \epsilon + \epsilon \sqrt{b-a}$.

Letting $\epsilon \to 0$, we have $||f||_{L^2(a,b)} = 0$, i. e. f(x) = 0, a. e. $x \in [a,b]$. The proof is completed.



解的存在性:



解的唯一性:



解的稳定性:



We show the instability of the inverse problem of dispersion relation by the special case. Taking $a=0,\ b=1,\ c=2,\ d=3,\ f_2(x)=f_1(x)+\sqrt{n}\cos(n\pi x)$, and $f_{1,2}$ are the solutions of $g_{1,2}$ with $g_i(y)=\int_0^1\frac{1}{y-x}f_i(x)dx$. As $n\to\infty$, it is obvious that

$$||f_2 - f_1||_{L^2(0,1)} = \left(\int_0^1 (\sqrt{n}\cos(n\pi x))^2 dx\right)^{1/2} = \frac{\sqrt{n}}{\sqrt{2}} \to \infty,$$

(3.7)

色散关系的 逆算子无界

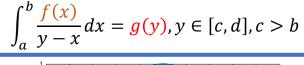
and

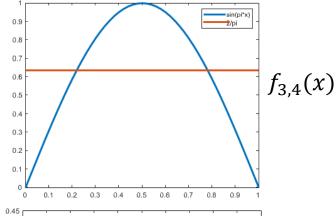
$$||g_2 - g_1||_{L^2(2,3)} = \frac{1}{\sqrt{n\pi}} \left(\int_2^3 \left(\int_0^1 \left(\frac{1}{y - x} \right)^2 \sin(n\pi x) dx \right)^2 dy \right)^{1/2} \le \frac{M}{\sqrt{n\pi}} \to 0.$$
 (3.8)

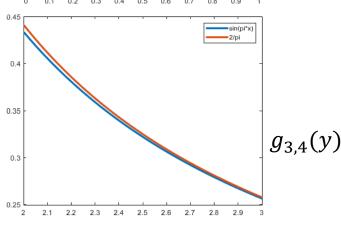
That means the solutions could be changed infinitely even though the noise of the input data is approaching to vanish. So the inverse problem is unstable.

实际物理中,数据一定存在误差:
$$\left|\left|g^{\delta}-g\right|\right|_{L^{2}}\leq\delta$$

不稳定性是求解该反问题的关键和难题!







正则化方法



目标:求解Kf = g

数据存在误差,只能求解 $Kf^{\delta}=g^{\delta}$,但 K^{-1} 无界,直接求解会导致 $f^{\delta}=K^{-1}g^{\delta}$ 不收敛到f。

$$\begin{aligned} \left| \left| f^{\delta} - f \right| \right|_{F} &\leq \left| \left| K^{-1} g^{\delta} - K^{-1} g \right| \right|_{F} \\ &\leq \left| \left| K^{-1} \right| \right| \left| \left| g^{\delta} - g \right| \right|_{G} = \left| \left| K^{-1} \right| \left| \delta \right| \end{aligned}$$

Andreas Kirsch.2011

正则化算子:

Define: A regularization strategy is a family of linear and bounded operators $R_{\alpha}: G \to F, \alpha > 0$, such that $\lim_{\alpha \to 0} R_{\alpha} K f = f$ for all $f \in F$, where the α is the regularization parameter [8].

构造一族有界算子 R_α 去逼近无界算子 K^{-1} ,即可将不适定问题转化成近似的适定问题。

$$f_{\alpha}^{\delta} = R_{\alpha}g^{\delta}$$

$$\begin{aligned} \left| \left| f_{\alpha}^{\delta} - f \right| \right|_{F} &\leq \left| \left| R_{\alpha} g^{\delta} - R_{\alpha} g \right| \right|_{F} + \left| \left| R_{\alpha} g - f \right| \right|_{F} & \delta \left| \left| R_{\alpha} \right| \right| \propto \frac{\delta}{\alpha^{M_{1}}}, M_{1} > 0 \\ &\leq \left| \left| R_{\alpha} \right| \right| \left| \left| g^{\delta} - g \right| \right|_{G} + \left| \left| R_{\alpha} K f - f \right| \right|_{F} & \left| \left| R_{\alpha} K f - f \right| \right|_{F} \propto \alpha^{M_{2}}, M_{2} > 0 \\ &\leq \left| \left| \left| R_{\alpha} \right| \left| \delta \right| + \left| \left| \left| R_{\alpha} K f - f \right| \right|_{F} \end{aligned} \right. \end{aligned}$$

严谨的正则化参数选取 方法均可选出合适的 α

正则化方法



Tikhonov正则化方法:

$$f_{\alpha}^{\delta} = argmin\left\{\frac{1}{2}||Kf - g^{\delta}||_{L^{2}}^{2} + \frac{\alpha}{2}||f||_{L^{2}}^{2}\right\}$$

Y. X. Zhang 2017

X. B. Yan 2018

X. B. Yan 2021

M. Asakawa 2001

第一项可看作χ²拟合, 第二项是罚项;

罚项的形式可根据实际问题做改进: H^1 、TV、 L^1 ······

 $\alpha > 0$ 是正则化参数,对结果有影响,不能过大或过小。

Tikhonov正则化对应的正则化算子: $R_{\alpha}=(K^*K+\alpha I)^{-1}K^*$, $f_{\alpha}^{\delta}=R_{\alpha}g^{\delta}$

Tikhonov正则化解和真解的估计:

先验信息:
$$f = K^*v, v \in G, ||v||_G \le E$$
 $||f_\alpha^\delta - f||_F \le \frac{\delta}{2\sqrt{\alpha}} + \frac{\sqrt{\alpha E}}{2}$

1、当误差趋于零时,正则化解一定能收敛到真解;

2、误差可以系统性控制。

正则化方法



正则化参数选取方法:

先验选取:结合先验信息 $f = K^*v$,理论上清楚展示正则化方法的收敛性;

后验选取:实际计算中方便运用,只知道输入数据或误差即可。

L-curve法:

 $\alpha = argmin \big\{ ||f_\alpha^\delta||_{L^2} ||Kf_\alpha^\delta - g^\delta||_{L^2} \big\}$

肖庭延 2003

 $||f_{\alpha}^{\delta}||_{\mathfrak{L}}$ 和 $||Kf_{\alpha}^{\delta}-g^{\delta}||$ 同时达到最小



Toy Models:

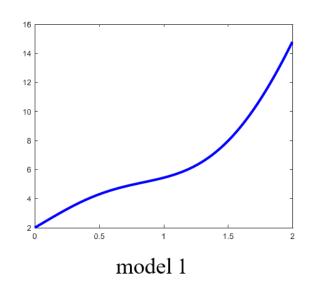
$$\int_{a}^{b} \frac{f(x)}{y-x} dx = g(y), y \in [c,d], c > b$$

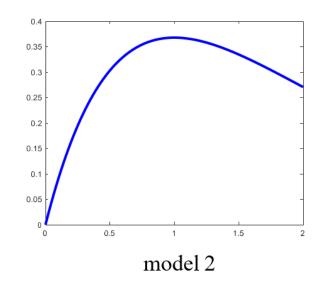
构造真实解 $f(x) = a_1 f_1(x) + a_2 f_2(x)$ 和对应的真实输入数据g(y),并对真实数据加误差得到误差数据 $g^{\delta}(y)$,以此来测试正则化方法的可靠性。

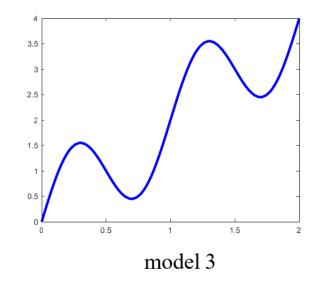
Model 1: a monotonic function as $f_1(x) = \sin(\pi x)$, $f_2(x) = e^x$;

Model 2: a simple non-monotonic function as $f_1(x) = xe^{-x}$, $f_2(x) = 0$;

Model 3: an oscillating function as $f_1(x) = \sin(2\pi x)$, $f_2(x) = x$.

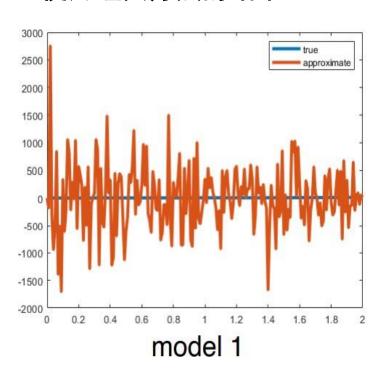


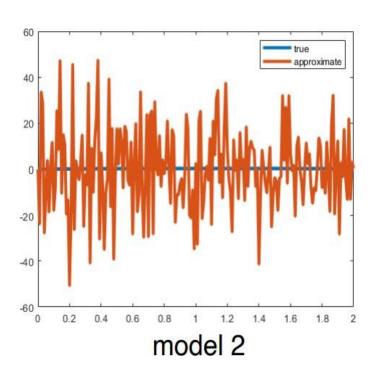


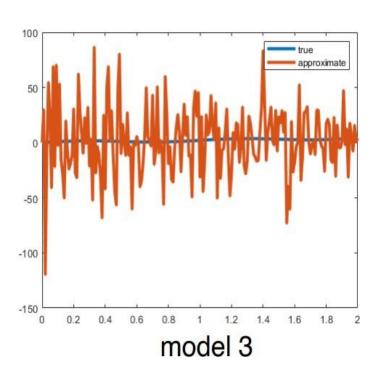




使用经典方法的结果:







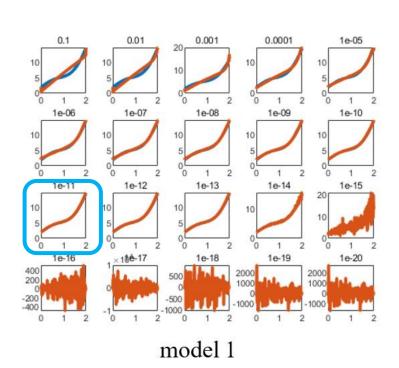
清楚地验证了该反问题的不稳定性;

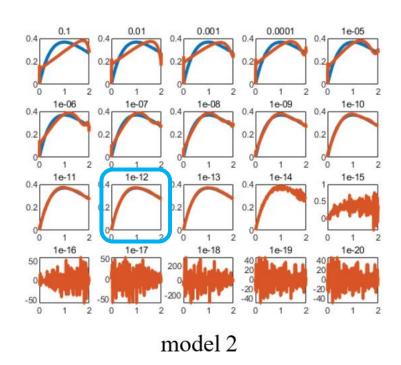
无法使用经典的方法求解, 必须要使用正则化方法。

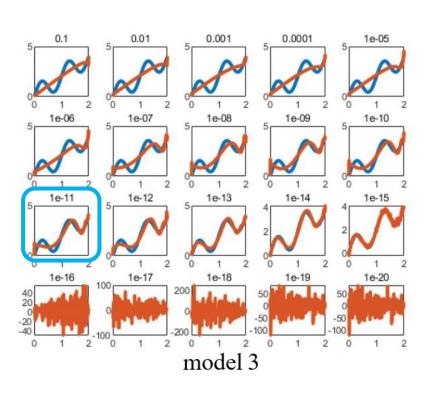




正则化参数对结果的影响:







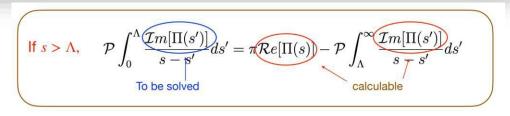
正则化参数 α 的选取很重要,符合理论预期;

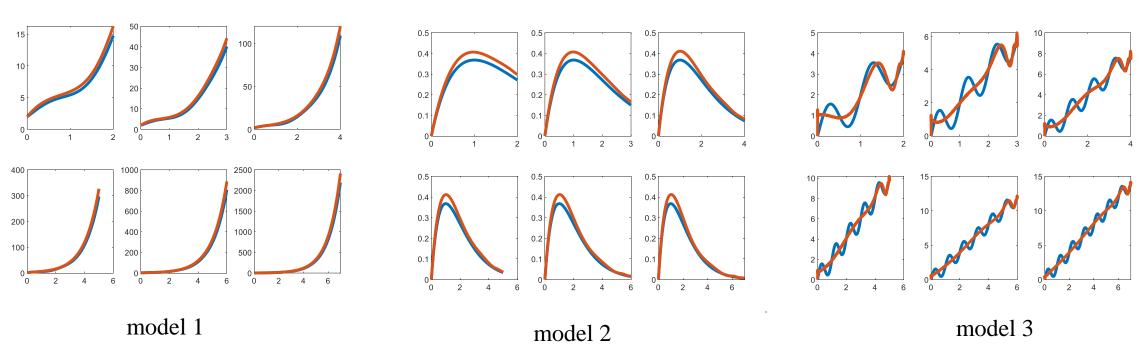
L-curve曲线法可以较好地选取α

正则化方法可以很好地解决色散关系的反问题。



截断 $\Lambda = \mathbf{b}$ 的敏感性: $\Lambda = 2, ..., 7$

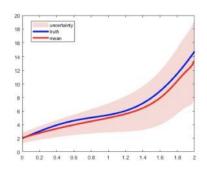


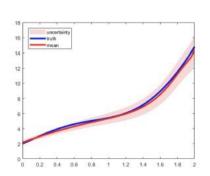


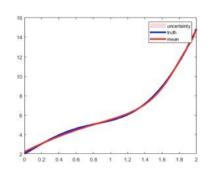
函数形式简单的函数, 截断Λ的影响不大。

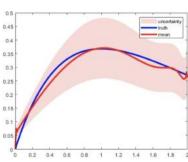


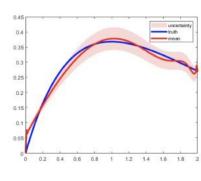
Tikhonov正则化方法和L-curve法的结果:

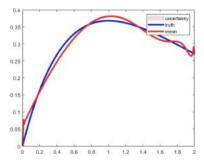


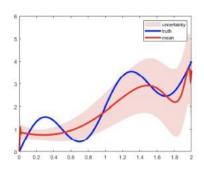


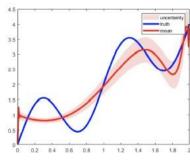


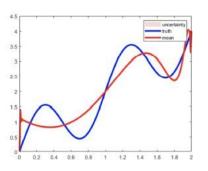












输入误差分别为30%, 10%, 1%;

随着输入误差降低,解的误差也逐渐降低,两者误差水平基本一致

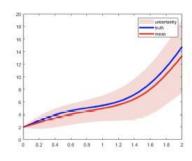
Model 1能够很好地反演,而Model 2和 Model 3的反演效果不是特别好;

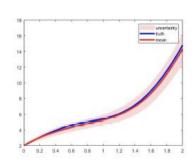
积分方程将解磨平

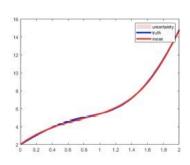
可对正则化方法进行改进。

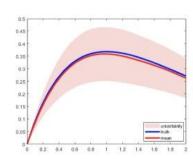


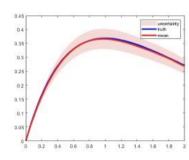
改进1: 限制解的空间: $||f||_{L^2}^2 \rightarrow ||f||_{H^1}^2$

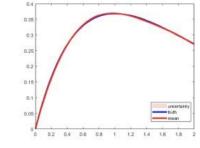


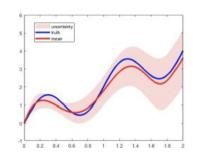


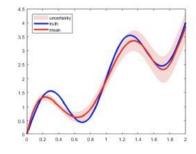


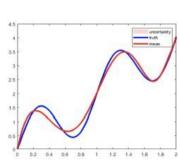












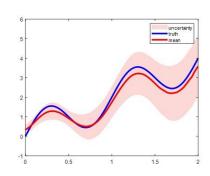
改进2:运用迭代,将每次反演的结果保存 并作为下次迭代的先验信息

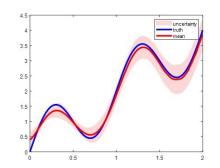
(1) Compute
$$r_k^{\delta} = g^{\delta} - K f_k$$

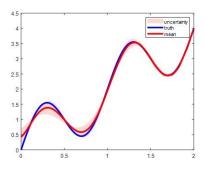
(2)
$$Solve \quad h_k = min\{\frac{1}{2}||Kh - r_k^{\delta}||_{L^2} + \frac{\alpha_k}{2}||h||_{H^1}\}$$

(3)
$$Update \quad f_{k+1} = f_k + h_k$$

(4) Stop by the L-curve method





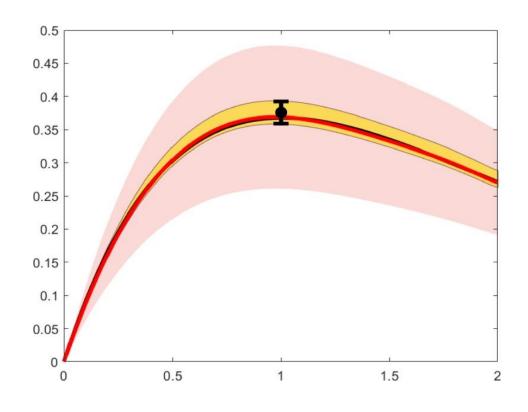




反问题方法的优势:

- 1. 基于严谨的数学框架;
- 2. 能够系统性地控制误差;
- 3. 不需要过大的计算资源;
- 4. 能够计算全部非微扰能区、激发态等;
- 5. 针对具体问题进行改进;
- 6. 可以与其他方法结合并互补;

与其他非微扰方法结合:





总结:

- 计算非微扰量的新方法;
- 严格证明色散关系反问题的不适定性;
- 运用正则化方法将不适定问题转化为适定性问题;
- 数值试验验证正则化方法的有效性。

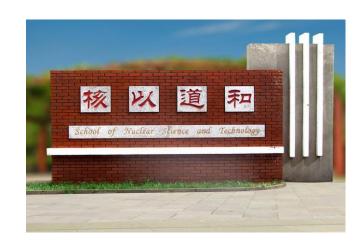
展望:

- 数学框架的进一步搭建;
- 物理问题的深入应用,解决重大基本问题;
- 寻找反问题方法运用的新方向。





谢谢大家的聆听!





熊傲昇(兰州大学) 2023年PQCD研讨会 2023年PQCD研讨会 22





- Solve $10^{-6}x = y$, $x, y \in R$, then $x = 10^6 y$.
- If y = 1, $y^{\delta} = 1.01$, $\delta = 0.01$, then $x = 10^6$, $x^{\delta} = 10^6 + 10^4$. (compute error is too large)
- Tikhonov regularized solution is

$$x_{\alpha}^{\delta} = 10^{-6} y^{\delta} / (\alpha + 10^{-12})$$

• If take $\alpha = 10^{-14}$, the $x_{\alpha}^{\delta} = 10^{6} = x$ (very good)





lpha	$\chi_{\scriptscriptstyle lpha}^{\scriptscriptstyle \delta}$
10^{-6} 10^{-7} 10^{-8} 10^{-9} 10^{-10} 10^{-11} 10^{-12} 10^{-13} 10^{-14} 10^{-15} 10^{-16} 10^{-17} 10^{-18} 10^{-18} 10^{-19}	1.009998990 10.099899001 100.989901010 1008.991008991 10000.000000000 91818.181818182 505000.0000000000 918181.818181818 100000.0000000000 1008991.008991009 1009989.900100999 1009998.990001010 1009999.899000010
10^{-20}	1009999.989900000

• How to choose α ?

In fact, we can prove that

$$\begin{aligned} \left| x_{\alpha}^{\delta} - x \right| &\leq \left| x_{\alpha}^{\delta} - x_{\alpha} \right| + \left| x_{\alpha} - x \right| \\ &\leq \left| \frac{(y^{\delta} - y)10^{-6}}{\alpha + 10^{-12}} \right| + \left| \frac{y10^{-6}}{\alpha + 10^{-12}} - x \right| \leq \frac{0.01}{2\sqrt{\alpha}} + \frac{|x|10^{6}\sqrt{\alpha}}{2} \end{aligned}$$

If we know $|x| \le 10^6$, let RHS=min i.e.

$$\frac{\sqrt{\alpha}}{2}10^{12} = \frac{\delta}{2\sqrt{\alpha}}$$

→

$$\alpha = \delta/10^{12} = 10^{-14}$$

We need to use some information of x which is unknown.