Inverse Problem Approach — A novel non-perturbative QCD method



Based on [Ao-Sheng Xiong(熊傲昇), Ting Wei(魏婷), FSY, arXiv:2211.13753]

2023.10.08, 中国科学院大学

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此方法还在发展过程中,此报告不是汇报结果而是介绍进展

希望通过报告听取大家更多的意见,让该方法发展得更加完善,

也希望更多人对该方法感兴趣,希望更多人参与发展和应用该方法。

Motivation: Problems of non-perturbation



- Particle physics: color confinement
- •New physics: muon g-2, Br(B)
- •Parton physics: mass and spin of nucleon, PDF, GPD, TMD, LCDA
- •Hadron physics: tetraquarks, pentaquark, glueballs
- •High energy nuclear physics: QCD phase transition, critical point
- •Low energy nuclear physics: nuclear force

$$B \to D^{(*)} \tau \nu) / Br(B \to D^{(*)} \ell \nu)$$



Motivation: non-perturbative approaches

- Lattice QCD
- •QCD sum rules
- Dyson-Schwinger Equation
- Chiral perturbation theory
- Holographic QCD
- Light-front quantization
- •Other EFTs and phenomenological models
- Each of them has its advantages and shortcomings.

• It is always welcome to develop a new theoretical method for non-perturbation,

to make complimentary predictions what are difficult by the above methods.

Criteria of a good theoretical approach

- (1) Well defined in mathematics
- (2) Realization in numerical calculations
- (3) Can be systematically improved
- (4) Simple at the beginning

The main idea of the inverse problem approach





- H.Umeeda, **FSY**, F.Xu, 2001.04079]
- Physical applications:
 - •muon g-2 [H.n.Li, H.Umeeda, 2004.06451]
 - •modifying the QCD sum rules [H.n.Li, H.Umeeda, 2006.16593]
 - •glueballs [H.n.Li, 2109.04956]
 - •pion distribution amplitudes [H.n.Li, 2205.06746]
 - •neutral meson mixings [H.n.Li, 2208.14798]
 - •understandings of fermion masses and EW masses [H.n.Li, 2302.01761, 2304.05921, 2306.03463]
- Its mathematical basis should be provided [A.S.Xiong, T.Wei, FSY, 2211.13753].

•Firstly proposed to solve the problem of understanding of $D^0 - \overline{D}^0$ mixing [H.n.Li,

The main idea of the inverse problem approach



- •With the dispersion relation of QFT, the non-perturbative quantities are obtained by solving the inverse problem with the perturbative calculations as inputs.
- •Using the regularization method, the solutions are stable, and can be converged to the true value as the input errors approaching zero.
- •The precision of the predictions can be systematically improved, without any artificially assumptions.

- 1. Dispersion relation and its inverse problem
- 2. Proof of ill-posedness
- 3. Regularization method
- 4. Test of some toy models
- 5. Physical discussions and perspectives

Outline

1. Dispersion relations and inverse problems

Dispersion relation:

- Based on Quantum Field Theory and correlation functions
- Analyticity of QFT, relation between a physical point and the curves, or relation between the real and imaginary parts

$$\Pi(q^2) = i \int d^4 x e^{iq \cdot x} \langle O(x)O(0) \rangle$$

$$Re[\Pi(s)] = \frac{1}{\pi} \int_0^\infty \frac{Im[\Pi(s')]}{s - s'} ds$$

• The above formula is just an example. Any dispersion relation would be studied similarly.



1. Dispersion relations and inverse problems





2. ill-posedness of the inverse problem

• $Kx = y \implies x = K^{-1}y$. Discretization?

$$\begin{cases} 2x_1 + 3x_2 = 5\\ 1.9999x_1 + 3.0001x_2 = 5 \end{cases}$$

$$\begin{cases} 2x_1 + 3x_2 = 5\\ 1.9999x_1 + 3.0001x_2 = 5.01 \end{cases}$$

•A very small noise might cause a large change of solutions



$$x_1 = -59, x_2 = 41$$

2. ill-posedness of the inverse problem

$$\begin{cases} 2x_1 + 3x_2 = 5\\ 1.9999x_1 + 3.0001x_2 = 5 \end{cases}$$
$$\begin{cases} 2x_1 + 3x_2 = 5\\ 1.9999x_1 + 3.0001x_2 = 5.01 \end{cases}$$

•A very small noise might cause a large change of solutions

$$K = \begin{pmatrix} 2 & 3\\ 1.9999 & 3.0001 \end{pmatrix}, \quad |K| = 0.0005, \quad K^{-1} = \frac{K^*}{|K|} = \begin{pmatrix} 6000.2 & -6000\\ -3999.8 & 4000 \end{pmatrix}$$
$$K^{-1} \text{ enhances the expression}$$

• In the continuum limit, K^{-1} is unbounded. The problem is ill-posed.

$$x_1 = 1, x_2 = 1$$

$$x_1 = -59, \ x_2 = 41$$



Dispersion relation: first-class Fredholm integration equation

If
$$s > \Lambda$$
, $\mathcal{P} \int_{0}^{\Lambda} \underbrace{\mathcal{I}m[\Pi(s')]}_{s - s'} ds' =$
To be solved

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \quad y \in \mathcal{S}_{a}^{b}$$

Existence of solution ? Uniqueness of the solution ?? Stability of the solution ???



$\in [c, d], c > b, a > 0$

ill-posedness of the inverse problem

- The operator $K: X \to Y$, Kx = y, X
- Inverse problem: solve x by known of K and y,
- Definition of well-posedness:

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 $f_n \to f(n \to \infty)$

- III-posedness: At least one of the above conditions is not satisfied
- If well-posed, K^{-1} must be a bounded or continuous operator, otherwise ill-posed.

Proof of uniqueness:

Proof. Since K is a linear operator, we know that just need to prove that Kf = 0 implies f(x) = 0, a

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 $\left|\frac{x}{y}\right| \le \left|\frac{b}{c}\right| < 1$, which implies that $\left|\sum_{k=0}^{\infty} \left(\frac{x}{y}\right)^k f(x)\right| \le \sum_{k=0}^{\infty} \left(\frac{b}{c}\right)^k |f(x)|$ 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

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

$$\int_{a}^{b} \frac{f(x)}{y-x} dx = g(y), \ y \in [c,d], \ c > b, \ a > 0$$

$$Kf_1 - Kf_2 = K(f_1 - f_2) = 0$$
. Setting $f = f_1 - f_2$, we
i. e. $x \in [a, b]$.

If $d < +\infty$, taking $z \in D := \{z \in \mathbb{C} : |z| \ge c\}$, we have

$$\Big|\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx\Big| \le \sum_{k=0}^{\infty} \frac{1}{c^k} |\int_a^b x^k f(x) dx| \le \sum_{k=0}^{\infty} \frac{b^k}{c^k} \int_a^b |f(x)| dx < +\infty,$$

which implies that the series $\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx$ is convergent uniformly on D. Since $\frac{1}{z^k} \int_a^b x^k f(x) dx$ is analytic on D for each k and use the Weierstrass theorem, we conclude that the series $\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx$ is analytic on *D*. Further, we know $\sum_{k=0}^{\infty} \frac{1}{v^k} \int_a^b x^k f(x) dx$ is real analytic on $y \in (c, +\infty)$. Combined with the analytic continuation, we know that (3.4) holds for y > c, i. e.

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

Similar to the proof process of the case $d = +\infty$, we also conclude that $\int_a^b x^k f(x) dx = 0, k = 0, 1, 2, \cdots$ for $d < +\infty$.



Proof of uniqueness:

Proof. Since *K* is a linear operator, we know that just need to prove that Kf = 0 implies f(x) = 0, a

Since C[a, b] is dense in $L^2(a, b)$, then for $f(x) \in L^2(a, b)$ and any $\epsilon > 0$, there exists $\tilde{f}(x) \in C[a, b]$, such that $||f - \tilde{f}||_{L^2(a,b)} < \epsilon$. On the other hand, for $\tilde{f}(x) \in C[a,b]$, there exists a polynomial $Q_n(x)$ of degree $n \in \mathbb{N}$, such that $\|\tilde{f} - Q_n\|_{C[a,b]} < \epsilon$ by the Weierstrass theorem. Therefore, we have

$$\begin{split} \|f - Q_n\|_{L^2(a,b)} &\leq \|f - \tilde{f}\|_{L^2(a,b)} + \|\tilde{f} - Q_n\|_{L^2(a,b)} \\ &\leq \epsilon + \sqrt{b - a} \|\tilde{f} - Q_n\|_{C[a,b]} \\ &< \epsilon + \epsilon \sqrt{b - a}, \end{split}$$

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \ y \in [c, d], \ c > b, \ a > 0$$

$$Kf_1 - Kf_2 = K(f_1 - f_2) = 0$$
. Setting $f = f_1 - f_2$, we
i. e. $x \in [a, b]$.

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

Proof of instability:

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

and

$$\|f_{2} - f_{1}\|_{L^{2}(0,1)} = \left(\int_{0}^{1} (\sqrt{n}\cos(n\pi x))^{2} dx\right)^{1/2} = \frac{\sqrt{n}}{\sqrt{2}} \to \infty,$$
(3.7)
$$\|g_{2} - g_{1}\|_{L^{2}(2,3)} = \frac{1}{\sqrt{n\pi}} \left(\int_{2}^{3} (\int_{0}^{1} (\frac{1}{y-x})^{2}\sin(n\pi x) dx)^{2} dy\right)^{1/2} \le \frac{1}{\sqrt{n\pi}} \to 0.$$
(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.

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \ y \in [c, d], \ c > b, \ a > 0$$

The inverse problem of dispersion relation is ill-posed See 2211.13753



Can we find a good solution? And how?

$$\frac{f(x)}{y-x} dx = g(y), \ y \in [c,d], \ c > b, \ a > 0$$

Inverse Problem Approach

- 1. Dispersion relation and its inverse problem
- 2. Proof of ill-posedness
- 3. Regularization method
- 4. Test of some toy models
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3. Regularization method

Define: such that $\lim_{\alpha \to 0} R_{\alpha} K f = f$ for all $f \in F$, where the α is the regularization parameter [8].

•Construct a bounded operator which is approximate to K^{-1} ,

- •III-posed problem => well-posed approximate problem, so that $f_{\alpha}^{\delta} = R_{\alpha}g^{\delta}$
- • f_{α}^{δ} is the approximate solution related to both α and δ .
- •An effective regularization strategy is to satisfy $f_{\alpha}^{\delta} \to f$, as $\|g^{\delta} g\| \leq \delta \to 0$

$$\begin{aligned} \left\| f_{\alpha}^{\delta} - f \right\|_{F} &\leq \left\| R_{\alpha} g^{\delta} - R_{\alpha} g \right\|_{F} + \left\| R_{\alpha} g - f \right\| \\ &\leq \left\| R_{\alpha} \right\| \left\| g^{\delta} - g \right\|_{G} + \left\| R_{\alpha} K f - f \right\| \\ &\leq \delta \left\| R_{\alpha} \right\| + \left\| R_{\alpha} K f - f \right\|_{F} \\ &\downarrow \qquad \downarrow \\ &\infty \qquad 0 \qquad \alpha \to 0 \end{aligned}$$

- A regularization strategy is a family of linear and bounded operators $R_{\alpha}: G \rightarrow F, \alpha > 0$,

- $\mathbf{R}_{\alpha}g f \|_{F}$ $Kf = g, f \in F, g \in G$
- $_{\alpha}Kf f\|_{F}$
- |F|

• To keep a balance, α can be neither too large nor too small

$$\lim_{\alpha \to 0} R_{\alpha} K f = f$$

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

$$R_{\alpha} := (\alpha I + K^* K)^{-1} K^*$$

$$f_{\alpha}^{\delta} = \arg\min_{f \in L^{2}(a,b)} J(f), \quad J(f) = \frac{1}{2} \|Kf - g^{\delta}\|_{L^{2}(c,d)}^{2} + \frac{\alpha}{2} \|f\|_{L^{2}(a,b)}^{2}$$

A priori condition: $f = K^*v, v \in G, ||v||_c$

Take $\alpha = \delta/E$

$$\|f_{\alpha}^{\delta} - f\|_{F} \leq \sqrt{\delta E} \to 0, \ \delta \to 0$$

/ Regularization

 $: G \to F \qquad \qquad \alpha f_{\alpha}^{\delta} + K^* K f_{\alpha}^{\delta} = K^* g^{\delta}$

$$_{G} \leq E \qquad \|f_{\alpha}^{\delta} - f\|_{F} \leq \frac{\delta}{2\sqrt{\alpha}} + \frac{\sqrt{\alpha}E}{2}$$

- •The most important: the uncertainty converges to vanishing as $\delta \rightarrow 0$.
- It exists an upper limit !
 The uncertainty must be controllable.

3. Selection rules of the Regularization parameter

A-priori methods are always difficult to use in practice. A-posterior methods can be tried.

 $\alpha = \arg\min_{f_{\alpha}^{\delta} \in L^{2}(a,b)} \left(\int_{a}^{\delta} f_{\alpha}^{\delta} f_{\alpha}^{\delta}$ L-curve method:

Both of $||f_{\alpha}^{\delta}||$ and $||g^{\delta} - Kf_{\alpha}^{\delta}||$ should be minimized together,

considering $f_{\alpha}^{\delta} = \arg \min J(f_{\alpha})$ $f \in L^2(a,b)$

$$\left(\left\|f_{\alpha}^{\delta}\right\|_{F}\left\|g^{\delta}-Kf_{\alpha}^{\delta}\right\|_{G}\right)$$

$$f), \quad J(f) = \frac{1}{2} \|Kf - g^{\delta}\|_{L^2(c,d)}^2 + \frac{\alpha}{2} \|f\|_{L^2(a,b)}^2$$

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4. Test of Toy Models

•Questions on the inverse problem approach:

- (1) **Regularization:** How important are the regularization methods? Can the solutions be systematically improved by the regularization method and the method of selecting the regularization parameter? (2) Impact of input uncertainties: What is the dependence of the errors of solutions
 - on the uncertainties of inputs? Larger, smaller or similar?
- (3) Impact of α and Λ : How sensitive are the solutions to the parameters α and Λ ? Does it exist a plateau?
- (4) Impact of more conditions: Can the solutions be improved if we known more conditions?

4. Test of Toy Models

- •Simple at the beginning: Tikhonov regularization + L-curve method for the regulator
 - in the future.

•Uncertainties are the most important issue.
$$b_i = \mu_i \pm \sigma_i$$

 $f(x) = a_1 f_1(x) + a_2 f_2(x)$ $g^{\delta}(y) = b_1 g_1(y) + b_2 g_2(y)$ $g_i(y) = \int_a^b \frac{f_i(x)}{y-x} dx$
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 $f_1(x) = \sin(2\pi x), f_2(x) = x$.

They are either helpful to clarify the properties of inverse problems or close to the real physical problem

• They are simple in mathematics and in practice and thus are very helpful to develop the new approach

0.5 1.0 1.5 2.0

4. Test: Importance of regularization

The solutions without any regularization:



- It can be clearly seen that the solutions are unstable and far from the true values.
- The ill-posed inverse problems can not be solved without any regularization.

nstable and far from the true values. Ived without any regularization.

4. Test: Importance of regularization

The solutions with Tikhonov regularization:



model 3



4. Test: Importance of regularization

The solutions with Tikhonov regularization:



- 2
- It can be seen clearly that some values of regularization parameters can give good results.
- The ill-posed inverse problems can be solved by regularization.
- The regularization parameter can be neither too small (not enough for regularization), nor too large (dominate over the original problem)
- $\mbox{-}$ But α still works by ranging several orders of magnitude.
- The regularization methods are very important in solving the inverse problems.

4. Test: Impact of input uncertainties

- The most important issue is to control the uncertainties!
 - The uncertainties of the solutions are almost at the same level of the input errors.
 - The smaller the input errors are, the more precise the solutions are.
 - The precision of the predictions can be systematically improved by lowering down the input errors.

Input errors:



30







4. Test: Impact of improved regularization method

- The regularization method can be modified according to the problem of physics
- The norm space of f(x) is changed from $L^2(a, b)$ to $H^1(a, b)$ $\|\|f\|_{L^2} = \int_a^b (f^2) dx \longrightarrow \|\|f\|_{H^1} = \int_a^b (f^2 + f'^2) dx$
 - The solutions are perfect for model 1 and 2. Model 3 is also significantly improved.
 - The uncertainties stemming from the regulator α is automatically included in the final results. Don't need to estimate the uncertainties from α . Input errors:















rs: 30%



1%

4. Test: Impact of improved regularization method

 The regularization method works well for the three models

model 2

- Non-stationary Tikhonov regularization for model 3
- (1) Compute $r_k^{\delta} = g^{\delta} K f_k$
- (2) Solve $h_k = min\{\frac{1}{2}||Kh r_k^{\delta}||_{L^2} + \frac{\alpha_k}{2}||h||_{H^1}\}$

model 3

- (3) *Update* $f_{k+1} = f_k + h_k$
- Stop by the L-curve method (4)

Input errors:



30%

10%

1%



4. Test: Plateaus of the regularization parameter α



There exist plateaus. Solutions are insensitive to regularization parameter. L-curve method is suitable. The inverse problem approach works for the non-perturbative calculations.



4. Test: Plateaus of the separation scale Λ



- There exist plateaus.
- The continuous condition at Λ might be even more helpful.

• Solutions are insensitive to the separation scale for monotonic and simple non-monotonic functions.



4. Test: Insensitivity to α and Λ



- Solutions are insensitive to the regularization parameter and the separation scale.
- The uncertainties of the inverse problem can be well controlled.



4. Test: Constrained data



•This method can combine with experiments and Lattice QCD to improve the precision of predictions

- Original uncertainty directly from inputs
- Data from experiments or Lattice QCD
- Improved uncertainty considering data

• If there is an experimental data or lattice data with much smaller uncertainty than the original solutions, we can use it to constrain the solution to be more precise in the whole range.






The precision can be systematically improved

- (1) Suitable regularization method and selection rule of the regulators
- (2) Higher precision of input data
- (3) Combination with higher precise data of experiments or Lattice QCD.

Without any beyond-control assumptions, the precision can be systematically improved:

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Criteria of a good theoretical approach

- (2) Realization in numerical calculations \longrightarrow Regularization methods

Inverse problem approach has a potential to be a first-principle approach

(1) Well defined in mathematics — Dispersion relation + proof of ill-posedness

- (3) Can be systematically improved \longrightarrow Errors converge to vanishing as $\delta \rightarrow 0$
- (4) Simple at the beginning Tikhonov regularization



5. Physical perspectives

- (1) Provide the quantities at the whole non-perturbative region
- (2) Advantage for the excited states. Either calculate directly, or combine with experiments or LQCD for ground states
- (3) Advantage for non-local correlation functions: widths and lifetimes, inclusive processes, distribution amplitudes
- (4) QCD sum rules with modification on the quark-hadron duality.
- (5) Solving some inverse problems in Lattice QCD.
- (6) More efforts on perturbative calculations to improve the input precision.
- (7) And many others...

$D^0 - \overline{D}^0$ Mixing

The time evolution

$$i\frac{\partial}{\partial t} \left(\begin{array}{c} D^{0}(t) \\ \overline{D}^{0}(t) \end{array} \right) = \left(\mathbf{M} - \frac{i}{2}\mathbf{\Gamma} \right) \left(\begin{array}{c} D^{0}(t) \\ \overline{D}^{0}(t) \end{array} \right)$$

Mixing parameters: Mass and Width differences •

$$x \equiv \frac{\Delta m}{\Gamma} = \frac{m_1 - m_2}{\Gamma}$$
 $y \equiv \frac{\Delta \Gamma}{2\Gamma} =$

- Useful to search for new physics, •
- but less understood in the Standard Model •











• After 2017, exclusive approach is dying

 $y_{\rm PP+PV} = (2.1 \pm 0.7) \times 10^{-3}$ Jiang, FSY, Qin, Li, Lü, '17



No theoretical methods work for D0 mixing No theoretical predictions for indirect CP violation

Inclusive Approach

Theory / Exp. comparison (for inclusive)



Hagelin 1981, Cheng 1982 Buras, Slominski and Steger 1984 NLO QCD Golowich and Petrov 2005



 $\begin{bmatrix} \mathrm{SM} \\ y \simeq 6 \times 10^{-7} \\ y \simeq 6 \times 10^{-7} \end{bmatrix}$

Suppressed by GIM

quark level

Short-distance

Exp.
$$\begin{cases} x = (3.9^{+1.1}_{-1.2}) \times 10^{-3} \\ y = (6.51^{+0.63}_{-0.69}) \times 10^{-3} \end{cases}$$

• For B_s, B_d mesons, the data are reproduced within 1σ .

• For D meson, the order of magnitude is not reproduced within leading-power.



 B_d meson

Artuso, Borissov and Lenz, 2016 Mruso, Borissov and Lenz, 2016 Mruso, Borissov and Lenz, 2016 Mruso, Borissov and Lenz, 2016 $\Delta M_s = (18.3 \pm 2.7) \text{ps}^{-1}$ Mruso, Borissov and Lenz, 2016 $\Delta M_d = (0.528 \pm 0.078) \text{ ps}^{-1}$ $\Delta \Gamma_d = (2.61 \pm 0.59) \cdot 10^{-3} \text{ ps}^{-1}$ HFLAV HFLAV HFLAV HFLAV HFLAV Mruso, Borissov and Lenz, 2016 $\Delta M_d = (0.5055 \pm 0.0020) \text{ ps}^{-1}$ $\Delta \Gamma_s = (0.082 \pm 0.006) \text{ ps}^{-1}$ $Exp. \begin{cases} \Delta M_d = (0.5055 \pm 0.0020) \text{ ps}^{-1} \\ \Delta \Gamma_d = 0.66(1 \pm 10) \cdot 10^{-3} \text{ ps}^{-1} \end{cases}$

Inverse Problem

$$D^0 - \overline{D}^0$$
 mixing $D^0 \stackrel{\longrightarrow}{\longrightarrow} \overline{D}^0$

$$\int_{0}^{\Lambda} ds' \frac{y(s')}{s-s'} = \pi x(s) - \int_{\Lambda}^{\infty} ds' \frac{y(s')}{s-s'} \equiv \omega(s)$$
parametrization:

$$y(s) = \frac{Ns[b_0 + b_1(s - m^2) + b_2(s - m^2)^2]}{[(s - m^2)^2 + d^2]^2}$$

Li, Umeeda, Xu, **FSY**, PLB(2020)





Additional conditions: data of x and y as inputs

Predict indirect CPV

 $q/p = 1.0002e^{i0.006^\circ}$

consistent with data $q/p = (0.969^{+0.050}_{-0.045})e^{i(-3.9^{+4.5}_{-4.6})^{\circ}}$





FIG. 7: Behaviors of x(s) (dotted line) and y(s) (solid line) for $\Lambda = 4.3 \text{ GeV}^2$.

 $x(m_D^2) = (0.21^{+0.04}_{-0.07})\%, \quad y(m_D^2)$ Inverse problem: $x = (0.44^{+0.13}_{-0.15})\%, \quad y = (0.63 \pm 0.07)\%,$ Experiment:

• Perspective: Using the Tikhonov regularization could provide more reasonable uncertainties.

A real prediction

$$n_D^2$$
) = (0.52 ± 0.03)%.

H.n.Li, 2208.14798



- Muon g-2: 4.2 σ deviation from the SM



- Inverse Problem:
- Result: Inverse problem: $a_{\mu}^{\text{HVP}} = (641^{+65}_{-63}) \times 10^{-10}$

Non-perturbative properties can be revealed from asymptotic QCD by solving an inverse problem.

Muon g-2, PRL(2021)

• Dominate uncertainty of the SM prediction: hadronic vacuum polarization (HVP) Aoyama, et al, Phys.Rept(2020)









• Perspective: 4-loop pQCD combined with experimental data at reliable regions might solve the BABAR-KLOE problem and lower down the uncertainty of predictions.







• Conventional QCD sum rules $\Pi_{\mu\nu}(q^2) = i \int$ Dispersion relation: $\Pi(q^2) = \frac{1}{2\pi i} \oint$ $\operatorname{Im}\Pi(q^2) = \pi f_V^2 d$ Quark-hadron duality: $ho^h(s) = rac{1}{\pi} {
m Im} \Pi^{
m I}$ $\int_{s_{h}}^{\infty} ds \frac{\rho^{n}(s)}{s - q^{2}} =$

• Uncertainty sources: quark-hadron duality. Results are very sensitive to the effective threshold s_0

QCD sum rules

$$\int d^4x e^{iq \cdot x} \langle 0|T[J_{\mu}(x)J_{\nu}(0)]|0\rangle$$

$$\int ds \frac{\Pi(s)}{s-q^2} = \frac{1}{\pi} \int_{t_{min}}^{\infty} ds \frac{\operatorname{Im} \Pi(s)}{s-q^2 - i\epsilon}$$

$$\delta(q^2 - m_V^2) + \pi \rho^h(q^2)\theta(q^2 - s_h)$$

$$\operatorname{Pert}(s)\theta(s-s_0)$$

$$= \frac{1}{\pi} \int_{s_0}^{\infty} ds \frac{\operatorname{Im} \Pi^{\operatorname{pert}}(s)}{s-q^2}$$

- Excited states and continuum spectrum can be directly solved by the inverse problem.
- Avoid the quark-hadron duality

$$\begin{split} \mathrm{Im}\Pi(q^2) &= \pi f_{\rho}^2 \delta(q^2 - m_{\rho}^2) + \pi f_{\rho(1450)}^2 \delta(q^2 - m_{\rho(1450)}^2) + \pi f_{\rho(1700)}^2 \delta(q^2 - m_{\rho(1700)}^2) \\ &+ \pi f_V^2 \delta(q^2 - m_V^2) + \pi \rho^h(q^2), \end{split}$$

H.n.Li, Umeeda, 2006.16593

$$m_{\rho(770)}(m_{\rho(1450)}, m_{\rho(1700)}, m_{\rho(1900)})$$

$$f_{\rho(770)}(f_{\rho(1450)}, f_{\rho(1700)}, f_{\rho(1900)}) \approx 0$$

• Perspective: Inverse problem modifies the QCD sum rules.

- Provide under-controlled uncertainties.
- Calculate whatever calculated.
- Advantage to excited states, no matter how much the pole contributes.

QCD sum rules

 $\approx 0.78 \ (1.46, 1.70, 1.90) \ \text{GeV}$

 $0.22 \ (0.19, \ 0.14, \ 0.14) \ \mathrm{GeV}$



Light-cone distribution amplitudes

- Theoretical uncertainties on baryon CPV are dominated by the baryon LCDAs.
- Limited knowledge for nucleons. VERY very limited for all the others, especially for HIGH TWISTs. $\phi_{\pi}(x)$
- LaMET and Lattice QCD

Z.F.Deng, C.Han, W.Wang, J.Zeng, J.L.Zhang, 2304.09004 Hua, et al, 2021

Inverse Problem can give very high moments.

 $(a_2^{\pi}, a_4^{\pi}, a_6^{\pi}, a_8^{\pi}, a_{10}^{\pi}, a_{12}^{\pi}, \cdots, a_{32}^{\pi}, a_{34}^{\pi})|_{\mu=2\,\mathrm{GeV}}$ $= (0.1775^{+0.0036}_{-0.0040}, 0.0957^{+0.0011}_{-0.0012}, 0.0762^{+0.0006}_{-0.0003}, 0.0688^{+0.0016}_{-0.0012}, 0.0643^{+0.0021}_{-0.0017}, 0.0603^{+0.0024}_{-0.0019$ $\cdots, 0.0089^{+0.0004}_{-0.0006}, 0.0028^{+0.0001}_{-0.0003}),$

• Perspective: Tikhonov regularization could provide more reasonable uncertainties.

H.n.Li, 2205.06746





1) 部分子分布 LCDA、PDF、GPD、碎裂函数 2) QCD相变 临界点位置、致密夸克物质

3) 强子物理与核物理

•激发态、多夸克态、核力

4) 新物理

•muon g-2、EDM等

Summary

- We propose a novel method to calculate the non-perturbative quantities.
- by solving the **inverse problem** with the perturbative calculations as inputs.
- artificial assumptions.
- The mathematical basis has been provided.
- Physical applications are expected.

• With the **dispersion relation** of QFT, the non-perturbative quantities are obtained

The precision of the predictions can be systematically improved, without any

Thank you!



Backups

后香农时代, 数学决定未来发展的边界

徐文伟 华为董事、战略研究院院长

MUAWEI

挑战问题7:反问题高精度快速求解

光存储在**密度、存储时间、成本和存储环境**要求上具备竞争力。尤其 是玻璃存储能够存储超千年。 挑战: 高密度要求多层和多通道, 不同层间或通道间的光干扰影响存储信号恢复的可靠性和精度。



数学模型

 $\mathbf{j}_{out} = \mathbf{J}_{Analyzer} \cdot \mathbf{J}_{Polarizer} \cdot \mathbf{J}_{sample} \cdot \mathbf{J}_{Polarizer} \cdot \mathbf{j}_{in}$

$$\mathcal{L} = \sum_{i=1}^{n} \left| \left| j_{i}(\delta, \theta) - \Phi(A_{i}, \Lambda) \right| \right|_{2}^{2} + \mathcal{R}$$
 正则化项

主要挑战

- 反问题中正则化方法的选取 ۰
- 层间相互干扰的模型构建
- 数值方法的稳定性
- 基于数据的模型修正策略
- 高效求解算法构造 •
- 算法与硬件的适配

问题:探索层间相互干扰和通道间相互干扰的模型, 寻找高精度、高速度、低延迟的算法,突破存 储的世界纪录

偏振光

Muller

Calculus

 $\begin{bmatrix} \exp[-i\phi_{1}^{+}] & 0 \\ 0 & \exp[-i\phi_{1}^{-}] \\ 0 & 0 & e \\ 0 & 0 & 0 \end{bmatrix}$

 $\begin{bmatrix} 0 & exp[-i\phi_{Z}^{*}] & 0 \\ 0 & 0 & exp[-i\phi_{Z}^{*}] \end{bmatrix}$

....

2. ill-posedness of the inverse problem

- The operator $K: X \to Y$, Kx = y, x
- The inverse problem of dispersion relation must be ill-posed.
- dimensional space.

Proof. It is easily to check that $Kf_1 + Kf_2 = K(f_1 + f_2)$ and $\alpha Kf = K(\alpha f)$ so the $K : F \to G$ operator is a linear operator. For any $f \in L^2(a, b)$, by the Cauchy inequality, we have

$$\begin{aligned} \|Kf\|_{L^{2}(c,d)}^{2} &= \int_{c}^{d} (Kf)^{2} dy = \int_{c}^{d} (\int_{a}^{b} \frac{1}{y-x} f(x) dx)^{2} dy \end{aligned}$$

$$\leq \int_{c}^{d} \int_{a}^{b} (\frac{1}{y-x})^{2} dx \int_{a}^{b} f^{2}(x) dx dy \leq (\frac{1}{c-b})^{2} (b-a) (d-c) \|f\|_{L^{2}(a,b)}^{2} = M \|f\|_{L^{2}(a,b)}^{2} < +\infty,$$
(3.2)

where M > 0 is a constant. Thus, from the form of the equation (3.2), we easily know $K : F \to G$ is a bounded operator.

Since c > b, the *m*th order derivative of *Kf* exists for any $m \in \mathbb{N}$ and by the Cauchy inequality, we have

$$\left\|\frac{\partial^m (Kf)}{\partial y^m}\right\|_{L^2(c,d)}^2 = \int_c^d (\int_a^b \frac{(-1)^m m!}{(y-x)^{m+1}} f(x) dx)^2 dy \le C \|f\|_{L^2(a,b)}^2, \tag{3.3}$$

where C > 0 is a constant depending on a, b, c, d only. Therefore, $Kf \in H^m(c, d)$ for any $m \in \mathbb{N}$. Since m is arbitrary, by the embedding theorem, we know $Kf \in C^{\infty}[c, d]$. And since $H^1(c, d)$ is embedded into $L^{2}(c, d)$ compactly, we know the operator K is a compact operator. The proof is completed

$$\in X, y \in Y$$

• K is a linear bounded compact operator. It doesn't have a bounded inverse operator in the infinite

2. Proof of the ill-posedness

Proof of uniqueness:

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

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

$$\int_{a}^{b} \frac{f(x)}{y-x} dx = g(y), \ y \in [c,d], \ c > b, \ a > 0$$

Proof. Since K is a linear operator, we know that $Kf_1 - Kf_2 = K(f_1 - f_2) = 0$. Setting $f = f_1 - f_2$, we

If $d < +\infty$, taking $z \in D := \{z \in \mathbb{C} : |z| \ge c\}$, we have

$$\Big|\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx\Big| \le \sum_{k=0}^{\infty} \frac{1}{c^k} |\int_a^b x^k f(x) dx| \le \sum_{k=0}^{\infty} \frac{b^k}{c^k} \int_a^b |f(x)| dx < +\infty,$$

which implies that the series $\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx$ is convergent uniformly on D. Since $\frac{1}{z^k} \int_a^b x^k f(x) dx$ is analytic on D for each k and use the Weierstrass theorem, we conclude that the series $\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx$ is analytic on D. Further, we know $\sum_{k=0}^{\infty} \frac{1}{v^k} \int_a^b x^k f(x) dx$ is real analytic on $y \in (c, +\infty)$. Combined with the analytic continuation, we know that (3.4) holds for y > c, i. e.

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

Similar to the proof process of the case $d = +\infty$, we also conclude that $\int_a^b x^k f(x) dx = 0, k = 0, 1, 2, \cdots$ for $d < +\infty$.



2. Proof of the ill-posedness

Proof of uniqueness:

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

By using (3.6), we know that $\int_{a}^{b} f(x)Q_{n}(x)dx = 0$. Combined with the Cauchy inequality, we have Since C[a, b] is dense in $L^2(a, b)$, then for $f(x) \in L^2(a, b)$ and any $\epsilon > 0$, there exists $\tilde{f}(x) \in C[a, b]$, $\|f\|_{L^{2}(a,b)}^{2} = \int_{a}^{b} f^{2}(x)dx = \int_{a}^{b} \left(f^{2}(x) - f(x)Q_{n}(x)\right)dx$ such that $||f - \tilde{f}||_{L^2(a,b)} < \epsilon$. On the other hand, for $\tilde{f}(x) \in C[a,b]$, there exists a polynomial $Q_n(x)$ of degree $n \in \mathbb{N}$, such that $\|\tilde{f} - Q_n\|_{C[a,b]} < \epsilon$ by the Weierstrass theorem. Therefore, we have $\leq \int_{a}^{b} |f(x)| \cdot |f(x) - Q_n(x)| dx$ $\leq \Big(\int_{a}^{b} f^{2}(x) dx \Big)^{\frac{1}{2}} \Big(\int_{a}^{b} |f(x) - Q_{n}(x)|^{2} dx \Big)^{\frac{1}{2}}$

$$\begin{split} \|f - Q_n\|_{L^2(a,b)} &\leq \|f - \tilde{f}\|_{L^2(a,b)} + \|\tilde{f} - Q_n\|_{L^2(a,b)} \\ &\leq \epsilon + \sqrt{b - a} \|\tilde{f} - Q_n\|_{C[a,b]} \\ &< \epsilon + \epsilon \sqrt{b - a}, \end{split}$$

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \ y \in [c, d], \ c > b, \ a > 0$$

$$= ||f||_{L^{2}(a,b)} ||f - Q_{n}||_{L^{2}(a,b)}$$
$$\leq (\epsilon + \epsilon \sqrt{b-a}) ||f||_{L^{2}(a,b)},$$

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

2. Proof of the ill-posedness

Proof of instability:

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

and

$$\|f_{2} - f_{1}\|_{L^{2}(0,1)} = \left(\int_{0}^{1} (\sqrt{n}\cos(n\pi x))^{2} dx\right)^{1/2} = \frac{\sqrt{n}}{\sqrt{2}} \to \infty,$$
(3.7)
$$\|g_{2} - g_{1}\|_{L^{2}(2,3)} = \frac{1}{\sqrt{n\pi}} \left(\int_{2}^{3} (\int_{0}^{1} (\frac{1}{y-x})^{2}\sin(n\pi x) dx)^{2} dy\right)^{1/2} \le \frac{1}{\sqrt{n\pi}} \to 0.$$
(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.

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \ y \in [c, d], \ c > b, \ a > 0$$

反问题是什么?

●反问题:

x? - 原因或输入

例:

- 小学期间 x和K是整数
- 中学期间 x是实数 K是映射
- 大学期间 x是向量 K是矩
 - x = x(t)是函数 K是
- 泛函: 算子方程 x是函数空间 K

→ F	$x \rightarrow y$	
过程或	乾模型 结果或输	〕出
	正问题	反问题
	求 $y = Kx$	$\boldsymbol{x} = \frac{1}{K} * \boldsymbol{y}$
封 或函数	求 $y = K(x)$	隐函数定理
巨阵	求 $y = K * x$	$\boldsymbol{x} = \boldsymbol{K^{-1}}\boldsymbol{y}$
是积分运算	求 $y(t) = \int \frac{x(s)}{t-s} ds$?
K是算子	$\mathbf{x}y = \mathbf{K}\mathbf{x}$	$x = K^{-1}y$

反问题是什么?

●反问题:

x? → 原因或输入

61.				
N3 •	常规一维函数	若 $y = ax + b(a \neq 0)$),已知 y_1 ,求 $x_1 = ?$	(存在且唯一)
		y = sin(x),	已知y ₂ ,求x ₂ =?	(存在但不唯一)
	矩阵问题	y = Kx,	已知y ₃ ,求x ₃ =?	依赖具体情况而定
		矩阵K非奇异		(存在且唯一)
	矩阵K奇异: 需要额外验证其他条件			(存在但不唯一) (不存在)



反问题是什么?

●反问题:





0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1









 $x(t) = t^2 + \sin(10\pi t)$







Proof of uniqueness:

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

It is easy to obtain that $Kf = \int_a^b \frac{1}{y-x} f(x) dx =$ c > b, we know $|\frac{x}{y}| \le |\frac{b}{c}| < 1$, which implies that 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

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

$$\int_{a}^{b} x^{k} f(x) dx =$$

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \ y \in [c, d], \ c > b, \ a > 0$$

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

$$t \left| \sum_{k=0}^{\infty} (\frac{x}{y})^{k} f(x) \right| \leq \sum_{k=0}^{\infty} (\frac{b}{c})^{k} |f(x)| \text{ for all } x \in [a, b].$$

$$0, \quad k = 0, 1, 2, \cdots. \tag{3.6}$$

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Proof of uniqueness:

If
$$d < +\infty$$
, taking $z \in D := \{z \in \mathbb{C} : |z| \ge c\}$, we have
$$\Big|\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx \Big| \le \sum_{k=0}^{\infty} \frac{1}{c^k} |\int_a^b x^k f(x) dx| \le \sum_{k=0}^{\infty} \frac{b^k}{c^k} \int_a^b |f(x)| dx < +\infty,$$

which implies that the series $\sum_{k=0}^{\infty} \frac{1}{z^k} \int_a^b x^k f(x) dx$ is convergent uniformly on D. Since $\frac{1}{z^k} \int_a^b x^k f(x) dx$ is analytic on D for each k and use the Weierstrass theorem, we conclude that the series $\sum_{k=0}^{\infty} \frac{1}{r^k} \int_a^b x^k f(x) dx$ is analytic on D. Further, we know $\sum_{k=0}^{\infty} \frac{1}{v^k} \int_a^b x^k f(x) dx$ is real analytic on $y \in (c, +\infty)$. Combined with the analytic continuation, we know that (3.4) holds for y > c, i. e.

$$\sum_{k=0}^{\infty} \frac{1}{y^k} \int_a^b x^k f(x)$$

Similar to the proof process of the case $d = +\infty$, we also conclude that $\int_a^b x^k f(x) dx = 0, k = 0, 1, 2, \cdots$ for $d < +\infty$.

 $x)dx = 0, \quad y \in (c, +\infty).$

Proof of uniqueness:

 $\|f\|_{L^2}^2$



Since C[a, b] is dense in $L^2(a, b)$, then for $f(x) \in L^2(a, b)$ and any $\epsilon > 0$, there exists $\tilde{f}(x) \in C[a, b]$, such that $||f - \tilde{f}||_{L^2(a,b)} < \epsilon$. On the other hand, for $\tilde{f}(x) \in C[a,b]$, there exists a polynomial $Q_n(x)$ of degree $n \in \mathbb{N}$, such that $\|\tilde{f} - Q_n\|_{C[a,b]} < \epsilon$ by the Weierstrass theorem. Therefore, we have

$$\begin{split} \|f - Q_n\|_{L^2(a,b)} &\leq \|f - \tilde{f}\|_{L^2(a,b)} + \|\tilde{f} - Q_n\|_{L^2(a,b)} \\ &\leq \epsilon + \sqrt{b-a} \|\tilde{f} - Q_n\|_{C[a,b]} \\ &< \epsilon + \epsilon \sqrt{b-a}, \end{split}$$

By using (3.6), we know that $\int_{a}^{b} f(x)Q_{n}(x)dx = 0$. Combined with the Cauchy inequality, we have

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

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.



Proof of instability:

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 $||f_2 - f_1||_{L^2(0,1)} = \left(\int_0^1 (\sqrt{1})\right)$ and $\|g_2 - g_1\|_{L^2(2,3)} = \frac{1}{\sqrt{n\pi}} \left(\int_2^3 (\int_0^1 (g_1 - g_2) g_2) g_2 (g_2 - g_2) g_1 \|_{L^2(2,3)} \right)$

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.

$$\int_{a}^{b} \frac{f(x)}{y - x} dx = g(y), \ y \in [c, d], \ c > b, \ a > 0$$

$$(\frac{1}{y-x})^{2} \sin(n\pi x) dx)^{2} dy \int^{1/2} \left(= \frac{\sqrt{n}}{\sqrt{2}} \to \infty, \right)$$
(3.7)
$$(\frac{1}{y-x})^{2} \sin(n\pi x) dx)^{2} dy \int^{1/2} \left(\le \frac{1}{\sqrt{n\pi}} \to 0, \right)$$
(3.8)

Numerical Method of Tikhonov Regularization

$$\varphi_{i}(x) = \begin{cases} \frac{x - x_{i-1}}{h}, x \in [x_{i-1}, x_{i}], \\ -\frac{x - x_{i+1}}{h}, x \in [x_{i}, x_{i+1}], \\ 0, otherwise, \end{cases}$$

$$f_{\alpha}^{\delta} = \underset{f \in L^{2}(a,b)}{\arg\min} J(f) = \underset{f \in L^{2}(a,b)}{\arg\min} \left(\frac{1}{2} \|Kf - g^{\delta}\|_{L^{2}(c,d)}^{2} + \frac{\alpha}{2} \|f\|_{L^{2}(a,b)}^{2} \right)$$

$$\varphi_0(x) = \begin{cases} -\frac{x-x_1}{h}, x \in [x_0, x_1], \\ 0, otherwise, \end{cases}$$

$$\begin{aligned} f^{\delta}_{\alpha,n}(x) &= \sum_{i=0}^{n} c_{i} \varphi_{i}(x) \\ & \left\| \sum_{i=0}^{n} c_{i} K \varphi_{i} - g^{\delta} \right\|_{L^{2}(c,d)}^{2} + \frac{\alpha}{2} \left\| \sum_{i=0}^{n} c_{i} \varphi_{i} \right\|_{L^{2}(a,b)}^{2} \\ c_{j}(K \varphi_{i}, K \varphi_{j})_{L^{2}(c,d)} - \sum_{i=0}^{n} c_{i}(K \varphi_{i}, g^{\delta})_{L^{2}(c,d)} + \frac{1}{2} (g^{\delta}, g^{\delta})_{L^{2}(c,d)} + \frac{\alpha}{2} \sum_{i,j=0}^{n} c_{i} c_{j} (\varphi_{i}, \varphi_{j})_{L^{2}(a,b)} \\ A_{ij} &= (K \varphi_{i}, K \varphi_{j})_{L^{2}(c,d)} \qquad B_{ij} = (\varphi_{i}, \varphi_{j})_{L^{2}(a,b)} \qquad C = (c_{0}, c_{1}, \cdots, c_{n})^{T} \end{aligned}$$

$$\varphi_n(x) = \begin{cases} \frac{x - x_{n-1}}{h}, x \in [x_{n-1}, x_n], \\ 0, otherwise. \end{cases}$$

$$\begin{split} f_{\alpha,n}^{\delta}(x) &= \sum_{i=0}^{n} c_{i} \varphi_{i}(x) \\ J(f_{\alpha,n}^{\delta}) &= \frac{1}{2} \left\| \sum_{i=0}^{n} c_{i} K \varphi_{i} - g^{\delta} \right\|_{L^{2}(c,d)}^{2} + \frac{\alpha}{2} \left\| \sum_{i=0}^{n} c_{i} \varphi_{i} \right\|_{L^{2}(a,b)}^{2} \\ &= \frac{1}{2} \sum_{i,j=0}^{n} c_{i} c_{j} (K \varphi_{i}, K \varphi_{j})_{L^{2}(c,d)} - \sum_{i=0}^{n} c_{i} (K \varphi_{i}, g^{\delta})_{L^{2}(c,d)} + \frac{1}{2} (g^{\delta}, g^{\delta})_{L^{2}(c,d)} + \frac{\alpha}{2} \sum_{i,j=0}^{n} c_{i} c_{j} (\varphi_{i}, \varphi_{j})_{L^{2}(a,d)} \\ &= A_{ij} = (K \varphi_{i}, K \varphi_{j})_{L^{2}(c,d)} \qquad B_{ij} = (\varphi_{i}, \varphi_{j})_{L^{2}(a,b)} \qquad C = (c_{0}, c_{1}, \cdots, c_{n})^{T} \end{split}$$

 $X_n = \operatorname{span}\{\varphi_0, \varphi_1, \cdots, \varphi_n\}$ $f_{\alpha,n}^{\delta}(x) = \sum_{i=0}^{n} c_i \varphi_i(x)$

 $D_i = (K\varphi_i, g^{\delta})_{L^2(c,d)}$ $(A + \alpha B)C = D$ **Theorem 4.3.** If the noise δ and the regularization parameter α are fixed, we have $||f_{\alpha,n}^{\delta} - f_{\alpha}^{\delta}||_{L^2(a,b)} \rightarrow$

0, $as n \to \infty$.



Dispersive analysis of neutral meson mixing

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- Revisited by the inverse matrix method
- •SU(3) breaking effects: physical thresholds of $D \rightarrow \pi \pi, K \pi, K K$
- •The solutions are stable
- •B mixing and kaon mixing are also studied in the same formalism

Inverse Problem: inverse matrix method

- arXiv:2109.04956 and on pion LCDA in arXiv:2205.06746
- A unique and stable solution can be attained before an ill posed nature appears. (Discretized regularization)



Notice that the range of $v = [0, +\infty)$

•The inverse matrix method was proposed by Hsiang-nan Li, on the studies of glueballs in



Inverse Problem: inverse matrix method

• Inverse Problem:
$$\int_0^\infty dv \frac{\Delta A_{ij}(v)}{u-v} = \int_0^\infty dv \frac{A_{ij}^{\text{box}}(v\Lambda + m_{IJ})e^{-v^2}}{u-v} + \int_{r_{ij}-r_{IJ}}^0 dv \frac{A_{ij}^{\text{box}}(v\Lambda + m_{IJ})}{u-v}$$

- •Expansion by generalized Laguerre polynomials: $\Delta A_{ij}(v) = \sum^{N}$ n=1
 - Unknown coefficients: $a^{(ij)} = (a_1^{(ij)})$

orthogonality:
$$\int_0^\infty v^\alpha e^{-v} L_m^{(\alpha)}(v) L_n^{(\alpha)}(v) dv = \frac{\Gamma(m+\alpha+1)}{m!} \delta_{mn} \quad \text{independent } \delta_{mn}$$

$$\sum_{n}^{(ij)} v^{\alpha} e^{-v} L_{n-1}^{(\alpha)}(v),$$

$$^{)},a_{2}^{(ij)},\cdots,a_{N}^{(ij)}),$$

 $\alpha = 3/2$ by physical condition of $\Delta A_{ij}(v) \sim v^{3/2}$ at $v \to 0$



• Inverse Problem:
$$\int_0^\infty dv \frac{\Delta A_{ij}(v)}{u-v} = \int_0^\infty dv \frac{A_{ij}^{\text{box}}(v\Lambda + m_{IJ})e^{-v^2}}{u-v} + \int_{r_{ij}-r_{IJ}}^0 dv \frac{A_{ij}^{\text{box}}(v\Lambda + m_$$

Expansion by generalized Laguerre polyn

$$U_{mn} = \int_{0}^{\infty} dv v^{m-1+\alpha} e^{-v} L_{n-1}^{(\alpha)}(v) \qquad \boxed{Ua^{(ij)} = b^{(ij)}}$$
$$b_{m}^{(ij)} = \int_{0}^{\infty} dv v^{m-1} A_{ij}^{\text{box}}(v\Lambda + m_{IJ}) e^{-v^{2}} + \int_{r_{ij}-r_{IJ}}^{0} dv v^{m-1} A_{ij}^{\text{box}}(v\Lambda + m_{IJ})$$

Inverse Problem: inverse matrix method

nomials:
$$\Delta A_{ij}(v) = \sum_{n=1}^{N} a_n^{(ij)} v^{\alpha} e^{-v} L_{n-1}^{(\alpha)}(v),$$



d
$$a^{(ij)} = U^{-1}b^{(ij)}$$

 $N \times N$: regularization



Inverse Problem: inverse matrix method

$$a^{(ij)} = U^{-1}b^{(ij)}$$

One can then solve for the vector $a^{(ij)}$ through $a^{(ij)} = U^{-1}b^{(ij)}$ by applying the inverse matrix U^{-1} . The existence of U^{-1} implies the uniqueness of the solution for $a^{(ij)}$. An inverse problem is usually ill posed; namely, some elements of U^{-1} rise fast with its dimension. Nevertheless, the convergence of Eq. (15) can be achieved at a finite N, before U^{-1} goes out of control. The difference between an obtained solution and a true one produces a correction to Eq. (14) only at power $1/u^{N+1}$, and the coefficients $a_n^{(ij)}$ built up previously are not altered by the inclusion of an additional higher-degree polynomial into the expansion in Eq. (15), because of the orthogonality condition in Eq. (16). The convergence of solutions in the polynomial expansion and their insensitivity to Λ will validate our approach, which is thus free of tunable parameters.
$M_{12}(s) = \frac{1}{2\pi} \int_{4\pi}^{\infty}$ Dispersion relation:

 $|D_{1,2}\rangle = p|D^0\rangle \pm q|\bar{D}^0\rangle \qquad \qquad \frac{q}{n} = \sqrt{\frac{2M_{12}^* - i\Gamma_{12}^*}{2M_{12} - i\Gamma_{12}}}$

$$x \equiv \frac{m_2 - m_1}{\Gamma} = \frac{1}{\Gamma} \operatorname{Re} \left[\frac{q}{p} (2M_{12} - i\Gamma_{12}) \right]$$

 $x = -\frac{2}{3}$ In the CP-conserving case:

•Absorptive piece: $\Gamma_{12}(s) = \sum_{i,j} \lambda_i \lambda_j \Gamma_{ij}(s)$ $i, j = d, s, b, \text{ and } \lambda_k \equiv V_{ck} V_{uk}^*, k = d, s, b$

 $D^0 - \overline{D}^0$ mixing

$$\sum_{n_{\pi}^{2}}^{\infty} ds' \frac{\Gamma_{12}(s')}{s-s'}$$





$$y \equiv \frac{\Gamma_2 - \Gamma_1}{2\Gamma} = -\frac{1}{\Gamma} \operatorname{Im} \left[\frac{q}{p} (2M_{12} - i\Gamma_{12}) \right]$$
$$\frac{2M_{12}}{\Gamma}, \quad y = \frac{\Gamma_{12}}{\Gamma}$$

 $\Gamma_{12}(m_D^2) = \lambda_s^2 [\Gamma_{dd}(m_D^2) - 2\Gamma_{ds}(m_D^2) + \Gamma_{ss}(m_D^2)] + 2\lambda_s \lambda_b [\Gamma_{dd}(m_D^2) - \Gamma_{ds}(m_D^2)] + \lambda_b^2 \Gamma_{dd}(m_D^2)$

 Dispersion relation: 	$M_{12}(s) = \frac{1}{2\pi} \int_{4m_{\tau}^2}^{\infty}$
 Absorptive piece: 	$\Gamma_{12}(s) = \sum_{i,j} \lambda_i \lambda_j \Gamma_{ij}(s)$
At large s	$\Gamma_{ij}^{\text{box}}(s) = \frac{G_F^2 f_D^2 m_W^3 I}{12\pi^2}$
	$egin{aligned} A_{ij}^{ ext{box}}(s) &= rac{\pi}{2x_D^{3/2}}rac{\sqrt{x_D^2}}{2x_D^3} \ & imes \left\{ \left(1+rac{x_i}{2} ight) ight\} ight\} \end{aligned}$

$$M_{12}(s) = \sum_{i,j} \lambda_i \lambda_j$$

$D^0 - \overline{D}^0$ mixing

 $i, j = d, s, b, \text{ and } \lambda_k \equiv V_{ck} V_{uk}^*, \ k = d, s, b$

$$\frac{A_D}{B_D} A_{ij}^{\mathrm{box}}(s)$$

$$\frac{x_D^2 - 2x_D(x_i + x_j) + (x_i - x_j)^2}{(1 - x_i)(1 - x_j)} \qquad \qquad x_i = m_i^2 / m_W^2 \qquad x_D = s / m_W^2$$

$$\frac{x_i x_j}{4} \left[3x_D^2 - x_D(x_i + x_j) - 2(x_i - x_j)^2 \right] + 2x_D(x_i + x_j)(x_i + x_j - x_D) \right\}$$

 $M_{ij}(s)$

 In principle, the dispersion relation, as a result of QCD dynamics which has nothing to do with the CKM factors, holds for each pair of the components Mij(s) and Fij(s) • Inverse problem for each components of ij. $A_{ij}(s)$ is monotonic, easily for solutions.

 $M_{12}(s) = \frac{1}{2\pi} \int_{4m^2}^{\infty}$ Dispersion relation:

$$A_{ij}^{\text{box}}(s) = \frac{\pi}{2x_D^{3/2}} \frac{\sqrt{x_D^2 - 2x_D(x_i + x_j) + (x_i - x_j)^2}}{(1 - x_i)(1 - x_j)} \\ \times \left\{ \left(1 + \frac{x_i x_j}{4} \right) [3x_D^2 - x_D(x_i + x_j) - 2(x_i - x_j)^2] + 2x_D(x_i + x_j)(x_i + x_j - x_D) \right\}$$

 $\Gamma_{ij}(s)$ grows like $s^{3/2}$, so the integration is divergent. Γ_{12} converges due to SU(3) cancellation.

Reformulate the disperimental

$$\Pi_{ij}(s) = M_{ij}(s) - i\Gamma_{ij}(s)/2$$
$$\frac{1}{2\pi i} \oint ds' \frac{\Pi_{ij}(s')}{s - s'} = 0$$
$$ds' \frac{\Gamma_{ij}(s')}{s - s'} + \frac{1}{2\pi i} \int_{C_R} ds' \frac{\Pi_{ij}^{\text{box}}(s')}{s - s'},$$

Persion relation:
$$\Pi_{ij}(s) = M_{ij}(s) - i\Gamma_{ij}(s)/2$$

 $\frac{1}{2\pi i} \oint ds' \frac{\Pi_{ij}(s')}{s-s'} = 0$
 $M_{ij}(s) = \frac{1}{2\pi} \int_{m_{IJ}}^{R} ds' \frac{\Gamma_{ij}(s')}{s-s'} + \frac{1}{2\pi i} \int_{C_R} ds' \frac{\Pi_{ij}^{\text{box}}(s')}{s-s'},$
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$D^0 - \overline{D}^0$ mixing

$$\sum_{m_{\pi}^{2}}^{\infty} ds' \frac{\Gamma_{12}(s')}{s-s'}$$





•Reformulate the dispersion relation: Π

 $M_{ij}(s) = \frac{1}{2\pi} \int_{m_{II}}^{R} ds$ Physical threshold:

Quark-level threshold: $M_{ij}^{\text{box}}(s) = \frac{1}{2\pi} \int_{m_{ij}}^{R}$

At large *s*, $M_{ij}(s) = M_{ij}^{box}(s)$, as heavy meson mixings.

$$\int_{m_{IJ}}^{R} ds' \frac{\Gamma_{ij}(s')}{s-s'} = \int_{m_{ij}}^{R} ds' \frac{\Gamma_{ij}^{\text{box}}(s')}{s-s'},$$

$D^0 - \overline{D}^0$ mixing

$$I_{ij}(s) = M_{ij}(s) - i\Gamma_{ij}(s)/2$$

 $\frac{1}{2\pi i} \oint ds' \frac{\Pi_{ij}(s')}{s-s'} = 0$

$$ds' rac{\Gamma_{ij}(s')}{s-s'} + rac{1}{2\pi i} \int_{C_R} ds' rac{\Pi^{\mathrm{box}}_{ij}(s')}{s-s'},$$



$$m_{\pi\pi} = 4m_{\pi}^2, \ m_{\pi K} = (m_{\pi} + m_K)^2, \ m_{\pi K}$$

$$ds' rac{\Gamma_{ij}^{\mathrm{box}}(s')}{s-s'} + rac{1}{2\pi i} \int_{C_R} ds' rac{\Pi_{ij}^{\mathrm{box}}(s')}{s-s'}, \qquad m_{dd} \in \mathcal{M}_{dd}$$

$$m_{dd} = 4m_d^2, \ m_{ds} = (m_d + m_s)^2,$$



• Reformulate the dispersion relation:

Introduce a subtracted unknown function:

$$\Delta\Gamma_{ij}(s,\Lambda) = \Gamma_{ij}(s) - \Gamma_{ij}^{\text{box}}(s)\{1 - \exp[-(s - m_{IJ})^2/\Lambda^2]\}$$

The scale Λ characterizes the order of s, at which $\Gamma_{ij}(s)$ transits to the perturbative expression $\Gamma_{ij}^{\text{box}}(s)$

Inverse problem:

$$\int_{m_{IJ}}^{\infty} ds' \frac{\Delta \Gamma_{ij}(s',\Lambda)}{s-s'} = \int_{m_{IJ}}^{\infty} ds' \frac{\Gamma_{ij}^{\text{box}}(s') \exp[-(s'-m_{IJ})^2/\Lambda^2]}{s-s'} + \int_{m_{ij}}^{m_{IJ}} ds' \frac{\Gamma_{ij}^{\text{box}}(s')}{s-s'}$$

$D^0 - \overline{D}^0$ mixing

$$\int_{m_{IJ}}^{R} ds \underbrace{\Gamma_{ij}(s')}_{s-s'} = \int_{m_{ij}}^{R} ds' \frac{\Gamma_{ij}^{\text{box}}(s')}{s-s'},$$

To be solved

Alternative formula, like $1 - \exp[-(s - m_{IJ})^3/\Lambda^3]$, only vary the solution by few percent The subtraction term can be regarded as an ultraviolet regulator.

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FIG. 1: Dependencies of $y_{ds}(s) \equiv \Gamma_{ds}(s)/\Gamma$ on s for N = 3 (dotted line), N = 8 (dashed line), N = 13 (solid line) and N = 23 (dot-dashed line) with $\Lambda = 5 \text{ GeV}^2$.

$$10^{5} \times (a_{1}^{(ds)}, a_{2}^{(ds)}, a_{3}^{(ds)}, \cdots, a_{12}^{(ds)}, a_{13}^{(ds)}, a_{14}^{(ds)}, \cdots, a_{22}^{(ds)}, a_{23}^{(ds)})$$

= $(4.04, 2.47, 1.45, \cdots, -2.08 \times 10^{-2}, -4.59 \times 10^{-3}, 9.25 \times 10^{-3}, \cdots, 7.49 \times 10^{-2}, 1.04),$



FIG. 2: Comparison of the solutions $y_{ij}(s) \equiv \Gamma_{ij}(s)/\Gamma$ (solid lines) with the inputs $y_{ij}^{\text{box}}(s) \equiv \Gamma_{ij}^{\text{box}}(s)/\Gamma$ (dashed lines) for (a) ij = dd, (b) ij = ds and (c) ij = ss at $\Lambda = 5 \text{ GeV}^2$.



FIG. 2: Comparison of the solutions $y_{ij}(s) \equiv \Gamma_{ij}(s)/\Gamma$ (solid lines) with the inputs $y_{ij}^{\text{box}}(s) \equiv \Gamma_{ij}^{\text{box}}(s)/\Gamma$ (dashed lines) for (a) ij = dd, (b) ij = ds and (c) ij = ss at $\Lambda = 5$ GeV².



FIG. 3: Dependencies of (a) $y_{dd} - 2y_{ds} + y_{ss}$, (b) $y_{dd} - y_{ds}$ and (c) $y_{dd}^{\text{box}} - 2y_{ds}^{\text{box}} + y_{ss}^{\text{box}}$ on s for $\Lambda = 5 \text{ GeV}^2$.



from left to right, in the cases (a) with and (b) without the second term in Eq. (20).



FIG. 5: Dependencies of $y(m_D^2)$ on Λ in the cases with (upper curve) and without (lower curve) the second term in Eq. (20).

FIG. 4: Solutions of y(s) for $\Lambda = 4.0 \text{ GeV}^2$, 4.5 GeV^2 , 5.0 GeV^2 and 5.5 GeV^2 , corresponding to the curves with the peaks



FIG. 2: Comparison of the solutions $y_{ij}(s) \equiv \Gamma_{ij}(s)/\Gamma$ (solid lines) with the inputs $y_{ij}^{\text{box}}(s) \equiv \Gamma_{ij}^{\text{box}}(s)/\Gamma$ (dashed lines) for (a) ij = dd, (b) ij = ds and (c) ij = ss at $\Lambda = 5 \text{ GeV}^2$.



FIG. 3: Dependencies of (a) $y_{dd} - 2y_{ds} + y_{ss}$, (b) $y_{dd} - y_{ds}$ and (c) $y_{dd}^{\text{box}} - 2y_{ds}^{\text{box}} + y_{ss}^{\text{box}}$ on s for $\Lambda = 5 \text{ GeV}^2$.

5. Physical applications: neutral meson mixing



$$\frac{(s-s_1)(s_1-s_2)(s_2-s)}{2\pi} \int_{s_{th}}^{\Lambda} \frac{\Gamma_{12}(s')}{(s'-s_1)(s'-s_2)}$$

$$= (s_1 - s_2)M_{12}(s) + (s_2 - s)M_{12}(s_1) + (s - s_1)M_{12}(s_1)$$

$$\frac{(s-s_1)(s_1-s_2)(s_2-s)}{2\pi} \int_{\Lambda}^{\infty} \frac{\Gamma_{12}(s')}{(s'-s)(s'-s_1$$



$$\Gamma_{21}^{q} = \frac{1}{2M_{B_{q}}} \operatorname{Disc} \langle \overline{B}_{q} | i \int d^{4}x \ T \left(\mathcal{H}_{eff}^{\Delta B=1}(x) \mathcal{H}_{eff}^{\Delta B=1}(0) \right) |$$



Inverse problem in Lattice QCD



Spectral function reconstruction from Euclidean lattices

Rothkopf, 2211.10680



Inverse problem in Lattice QCD

Hadronic on the Lattice

Lattice QCD: Euclidean field theory using the path-integral formalism: time-dependent matrix elements are problematic.

$$W_{\mu\nu} = \frac{1}{4\pi} \int d^4 z e^{iq \cdot z} \left\langle p, s \left| \left[J^{\dagger}_{\mu}(z) J_{\nu}(0) \right] \right| p, s \right\rangle$$

Euclidean hadronic tensor:

$$\tilde{W}_{\mu\nu}(\vec{p},\vec{q},\tau=t_2-t_1) = \sum_{\vec{x}_2 \neq 1} e^{-i\vec{q} \cdot (\vec{x}_2 - \vec{x}_1)} \langle p, s | J^{\dagger}_{\mu}(\vec{x}_2,t_2) J_{\nu}(\vec{x}_1,t_1) | p, s \rangle$$

Back to Minkowski space by solving the inverse problem:

 $ilde{W}_{\mu
u}(\boldsymbol{p},\boldsymbol{q}, au)$

$$f(\boldsymbol{p}) = \int d\nu W_{\mu\nu}(\boldsymbol{p}, \boldsymbol{q}, \nu) e^{-\nu\tau}$$

K.F. Liu and S. J. Dong, PRL 72, 1790 (1994) K.-F. Liu, PRD62, 074501 (2000) J. Liang et. al., PRD101, 114503 (2020) J. Liang et. al., PRD102, 034514 (2020)

Jian Liang's talk @ 2nd EicC CDR workshop



Maximum entropy method (MEM)

- number of noisy data.
- Its basis is Bayes' Theorem:

 $P[X|Y] = \frac{P[Y|X]P[X]}{P[Y]}$ $P[A|DH] = \frac{P[D|AH]P[A|H]}{P[D|H]}.$ The most probable image is A(w) that satisfies the condition: $\frac{\delta P[A|DH]}{\delta A} = 0.$

From Bayes' Theorem, we can get :

(1) Firstly, they make:

 $\chi^2 - fitting$ does not work.

$$D(\tau) = \int_0^\infty K(\tau, w) A(w) \, dw$$

MEM is a method to circumvent these difficulties by making a statistical inference of the most probable SPF (or sometimes called the image in the following) as well as its reliability on the basis of a limited

 $D_A(\tau_i))C_{ij}^{-1}(D(\tau_j) - D_A(\tau_j)),$

In the case where P[A|H] = 0, maximizing P[A|DH] is equivalent to standard $\chi^2 - fitting$. However, the

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