# Some approaches to tackle ill-posed inversion problem

Hai-Tao Shu

#### Central China Normal University





## Outline



- Introduction & Motivation
- Methods
- Mock data tests
- Real lattice results

## **Difficulties in the inversion**

Relation between correlators and spectral functions:

$$G(\tau, \vec{p}, T) = \int_{0}^{\infty} \frac{d\omega}{2\pi} \rho(\omega, \vec{p}, T) K(\tau, \omega, T)$$

$$K(\tau, \omega, T) = \frac{\cosh(\omega(\tau - 1/2T))}{\sinh(\omega/2T)}$$

$$K(\tau, \omega, T) = \frac{\cosh(\omega(\tau - 1/2T))}{\sinh(\omega/2T)}$$

Spectral functions are crucial to understand in-medium hadron properties and transport properties of QGP:

$$G(\tau, \boldsymbol{p}) = \int d\boldsymbol{x} \, e^{-i \cdot \boldsymbol{p} \cdot \boldsymbol{x}} \langle J_H(0, 0) J_H^{+}(\tau, \boldsymbol{x}) \rangle$$

$$J_{H}(\tau, \vec{x}) = \overline{q}(\tau, \vec{x})\Gamma_{H}q(\tau, \vec{x})$$
  
Heavy quark diffusion coefficient
$$D = \frac{\pi}{{}^{6}\chi_{00}} \lim_{\omega \to 0} \frac{\rho_{ii}(\omega, \vec{p} = 0, T)}{\omega}$$

 $J_{H}(\tau, \vec{x}) = T^{\mu\nu}(\tau, \vec{x})$ Shear viscosity:  $\eta = \pi \underset{\omega \to 0}{lim} \underset{p \to 0}{lim} \frac{\rho^{12,12}(\omega, p, T)}{\omega}$ 

## **Maximum Entropy Method(MEM)**



Goal: obtain the most probable solution

$$P[\alpha|\bar{G}] = \frac{P[\alpha]}{P[\bar{G}]} \int \mathcal{D}\rho P[\bar{G}|\rho,\alpha] \cdot P[\rho|\alpha] \qquad \text{sharp-peak} \\ P[\bar{G}|\rho,\alpha] \sim \exp(-\chi^2[\rho]/2): \text{ likelihood function} \qquad \text{assumption} \\ P[\rho|\alpha] \sim \exp(\alpha S[\rho]): \text{ prior probability} \qquad \text{minimize } F = \frac{\chi^2}{2} - \alpha S \\ S[\rho] = \int d\omega \left[ \rho(\omega) - D(\omega) - \rho(\omega) \ln \frac{\rho(\omega)}{D(\omega)} \right] \\ P[\rho|\alpha] = \int d\omega \left[ \rho(\omega) - D(\omega) - \rho(\omega) \ln \frac{\rho(\omega)}{D(\omega)} \right]$$

[M. Jarrell, J. E. Gubernatis, Phy. Rep.269,133(1996)] [M. Asakawa et al., Prog.Part.Nucl.Phys. 46(2001) 459-508]

## Stochastic Analytic Inference(SAI)

A stochastic method based on Bayesian theorem :  $\langle n \rangle = \int d\alpha \, \langle n \rangle_{\alpha} \, P[\alpha | \overline{G}]$ 

[H. Ohno, PoS(LATTICE 2015)175]

Goal: find the distribution of  $P[\alpha | \overline{G}]$ 

- Field treatment of n(x) gives:  $\langle n \rangle_{\alpha} = \int \mathcal{D}n \, n \, P[n|\alpha, \overline{G}] = \int \mathcal{D}n \, n \frac{1}{Z(\alpha)} e^{-\chi^2/2\alpha}$ Posterior probability:  $P[n|\alpha, \overline{G}] = \frac{1}{P[\overline{G}|\alpha]} P[\overline{G}|\alpha, n] P[n|\alpha]$   $\begin{cases} P[\overline{G}|\alpha, n] = \frac{1}{Z'} e^{-\chi^2[n]/2\alpha}, \text{likelihood function} \\ P[n|\alpha] = \Theta(n)\delta\left(\int_{0}^{x_{max}} dx \, n(x) 1\right), \text{ prior probability} \\ P[\overline{G}|\alpha] = Z/Z', \text{ normalization factor} \end{cases}$
- $\geq P[\alpha|\bar{G}] = \frac{P[\alpha]}{P[\bar{G}]} \int \mathcal{D}n \ P[\bar{G}|\alpha, n] P[n|\alpha] \sim P[\alpha] \alpha^{-\frac{N}{2}} Z(\alpha)$



## **Stochastic Analytic Inference**



> Density of States(DoS):  $\Omega(E) = \int \mathcal{D}n \,\delta(\chi^2[n]/2 - E)$ >  $P[\alpha|\bar{G}] = P[\alpha]\alpha^{-\frac{N}{2}} \int dE \Omega(E)e^{-E/\alpha}$ 

F.-G. Wang, D.P. Landau arXiv:cond-mat/0107006



## **Stochastic Optimization Method(SOM)**

1e14

d<sup>2</sup>log(E)/d<sup>2</sup>log

[H.-T. Shu, H.-T. Ding, O. Kaczmarek, S. Mukherjee, H. Ohno, PoS(LATTICE 2015)180]

 Based on Central Limit Theorem.
 No prior information is needed. All information comes from correlators:

 $E = \chi^2[\rho]/2$  (Fictitious energy)

- > Field treatment of  $\rho$ . Evolves with fictitious temperature  $\alpha$ .
- Possible solution obtained when phase transition occurs.

[K.S.D. Beach arXiv:cond-mat/0403055]



ł²log(E)/d²log(α)

Phase transition occurs at the *kink(black spot)*.

## SAI to SOM

**D** SAI reduces to SOM when using **constant default model**!  $x = \omega, n(x) = \rho(\omega)$ 

## **Basis of Stochastic approaches**





#### SOM



#### **Mock data test**:Different spectral functions





SOM gives similar results to SAI with constant DM.
 SAI&SOM reconstruct the input well.

### **Mock data test**:Different spectral functions





1. MEM&SAI with constant DM and SOM can not reconstruct the transport peak precisely.

2. MEM&SAI&SOM reconstruct the resonance peak, but the width and peaklocation differ from the input.

3. SAI&SOM can reconstruct the continuum part well.

#### **Mock data test**: Dependence on $N_{\tau}$





1. MEM reconstructs the resonance peak well even at small  $N_{\tau}.$ 

2. SAI&SOM give fake transport peak at small  $N_{\tau}$  .

3. SAI&SOM reconstruct the continuum part well.

4. MEM gives fake resonance peaks.

#### **Mock data test**: Dependence on noise level





 MEM gives bad output with noisy data.
 SAI&SOM give a rough resonance peak with noisy data.

3. SAI&SOM reconstruct the continuum part well.

4. As noise becomes weak, the output approaches to the input for all methods.



1. Dependence is weak for resonance peak and continuum.

2. There is dependence for transport peak. But upper bound exists in this case.

#### **Real Lattice Results at 0.75Tc & 1.5Tc**



14



**BACKUP PAGES** 



α

#### BACKUP PAGES Mock SPFs



$$\geq \text{Resonance peak} : \rho_{res} = c_{res} \frac{\Gamma(\omega, \omega_0, \gamma_0)M}{(\omega^2 - M^2)^2 + M^2 \Gamma^2(\omega, \omega_0, \gamma_0)} \frac{\omega^2}{\pi}$$

$$\text{Where } \Gamma(\omega, \omega_0, \gamma_0) = \theta(\omega - \omega_0)\gamma_0(1 - \frac{\omega^2_0}{\omega^2})^5$$

Transport peak:

$$\rho_{trans} = c_{trans} \frac{\eta \omega}{(\omega^2 - \eta^2)^2}$$

Free continuum:

$$\rho_{cont} = c_{cont} \frac{N_c}{8\pi^2} \Theta(\omega^2 - 4m^2)\omega^2 \tanh\left(\frac{\omega}{4T}\right) \\ \times \sqrt{1 - \left(\frac{2m}{\omega}\right)^2} \left[a^1 + a^2\left(\frac{2m}{\omega}\right)^2\right]$$

Free Wilson:

$$\rho_{Wilson} = c_{Wilson} \frac{N}{L^3} \sum_k \sinh\left(\frac{\omega}{2T}\right) \left[b^1 - b^2 \frac{\left(\sum_{i=1}^3 \sin^2 k_i\right)}{\sinh^2 E_k(m)}\right]$$
  
Where  $\cosh E_k(m) = 1 + \frac{K_k^2 + M_k^2(m)}{2(1+M_k(m))}$ ,  $K_k = \sum_{i=1}^3 \gamma_i \sinh k_i$ 

#### **BACKUP PAGES** Mock SPFs parameters



Elements in SPF	Parameters
$ ho_{res}^1$	$c_{res} = 1, \omega_0 = 0.2, \gamma_0 = 0.20, M = 0.5$
$ ho_{res}^2$	$c_{res} = 4, \omega_0 = 0.2, \gamma_0 = 0.25, M = 1.2$
$ ho_{res}^3$	$c_{res} = 6, \omega_0 = 0.2, \gamma_0 = 0.20, M = 2.5$
ρ <sub>trans</sub>	$c_{trans} = 0.2, \eta = 0.01$
$ ho_{cont}$	$c_{cont} = 20, a^1 = 2, a^2 = 1, m = 0.1$
<i>ρ</i> <sub>Wilson</sub>	$c_{Wilson} = 1, b^1 = 3, b^2 = 1, m = 0.112$

 $N_{\tau} = 48, a = 1, \tau_{min} = 1, \omega \in [0,4].$ Error in mock data:  $\sigma(\tau) = \epsilon G(\tau)\tau$