

# Introduction to $\Omega$ *MaxEnt*, a tool for analytic continuation of Matsubara data

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# Analytic Continuation

- $G(\tau)$  or  $G(i\omega_n) \Rightarrow A(\omega)$ ?

$\Rightarrow$  invert

$$G(\tau) = - \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} \frac{e^{-\omega\tau} A(\omega)}{e^{-\beta\omega} + 1} .$$

or

$$G(i\omega_n) = \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} \frac{A(\omega)}{i\omega_n - \omega} .$$

# Conditioning Problem:

- If we discretize  $\omega$ :  $G = KA \Rightarrow A = K^{-1}G$
- error on  $A$ :

$$\frac{1}{\|K\| \|K^{-1}\|} \frac{\|\delta G\|}{\|G\|} \leq \frac{\|\delta A\|}{\|A\|} \leq \|K\| \|K^{-1}\| \frac{\|\delta G\|}{\|G\|}$$

- $\|K\| \|K^{-1}\|$  is large  $\Rightarrow \frac{\|\delta A\|}{\|A\|}$  not bounded
- Analytic continuation unique in principle (Baym and Mermin, J.Math.Phys.1961), but unstable numerically

$\Rightarrow$  need constraints on  $A(\omega)$

# Maximum entropy

- Different strategy: minimize

$$Q = \chi^2 - \alpha S$$

$$\chi^2 = \sum_{mn} (G_m - K_m A)^T C_{mn}^{-1} (G_n - K_n A)$$

$$S = - \int d\omega A(\omega) \ln \frac{A(\omega)}{D(\omega)}$$

$$C_{mn} = \langle (G_m - \langle G_m \rangle)(G_n - \langle G_n \rangle) \rangle$$

# $\Omega$ MaxEnt

- Use 
$$G(i\omega_n) = \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} \frac{A(\omega)}{i\omega_n - \omega} .$$

if  $A(\omega)$  is a piecewise polynomial  $\Rightarrow$  analytical integration in intervals

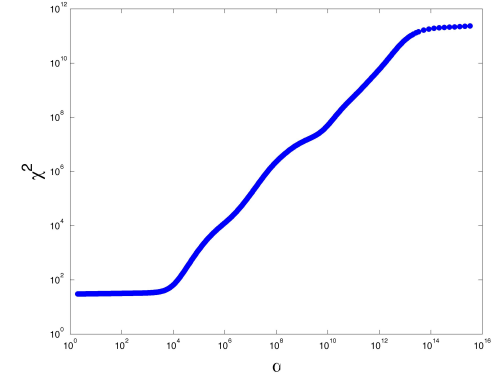
- Replace high Matsubara frequencies with constraints on moments

$$M_j = \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} \omega^j A(\omega)$$

- Use adapted  $\omega$  grid
- Input can also be  $G(\tau)$
- Treats fermionic ( $A(\omega) > 0$ ) and bosonic ( $A(\omega)/\omega > 0$ ) data
- General covariance matrix

# $\Omega$ MaxEnt: how to choose $\alpha$ ?

- Compute  $A(\omega)$  for large range of  $\alpha$  :
  - Three regimes in  $\chi^2$  vs  $\alpha$   
 $\Rightarrow$  Optimal  $\alpha$  located on  $\chi^2$  vs  $\alpha$  in *log-log*
- Additional diagnostic tools to assess quality of the result:
  - $A(\omega_{sample})$  vs  $\alpha$
  - $\Delta G = (G_{in} - G_{out}) / \sigma$  vs  $\omega_n$
  - $\langle \Delta G_m \Delta G_{m+n} \rangle$



# Why three regimes in $\chi^2$ vs $\alpha$ ?

$$Q = \chi^2 - \alpha S \quad \nabla_A Q = 0$$

- Large  $\alpha$ :  $A(\omega) \approx D(\omega) \Rightarrow \chi^2 \approx \text{const}$
- Intermediate:  $\alpha \searrow \Rightarrow \chi^2 \searrow$
- Small  $\alpha$ :  $\chi^2 \searrow$  very slowly (why?)

$$\alpha_j = \alpha_{j-1} - \Delta\alpha \quad \Rightarrow \quad A_j = A_{j-1} + \delta A_j$$

$$\delta A_j = \frac{\Delta\alpha}{2} \left[ \tilde{\mathbf{K}}^T \tilde{\mathbf{K}} + \frac{\alpha_j}{2} \Delta\omega \mathbf{A}_{j-1}^{-1} \right]^{-1} (\Delta\omega \ln(\mathbf{D}^{-1} A_{j-1}) + d\omega)$$

All quantities are smooth in RHS at optimal  $\alpha$  ,  
but smooth  $\delta A_j$  cannot fit noise

# Thanks!

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