

# A Parameter Free Genetic Algorithm for Estimating the Dynamic Structure Factor at Zero and Finite Temperature

**Nathan Nichols**  
(University of Vermont)

Adrian Del Maestro  
(University of Vermont)

Timothy Prisk  
(National Institute of  
Standards and Technology)

Garfield T Warren  
(Indiana University  
Bloomington)

Paul E Sokol  
(Indiana University  
Bloomington)

DMR-1809027

DMR-1808440

OAC-1827314



The University of Vermont

# The problem

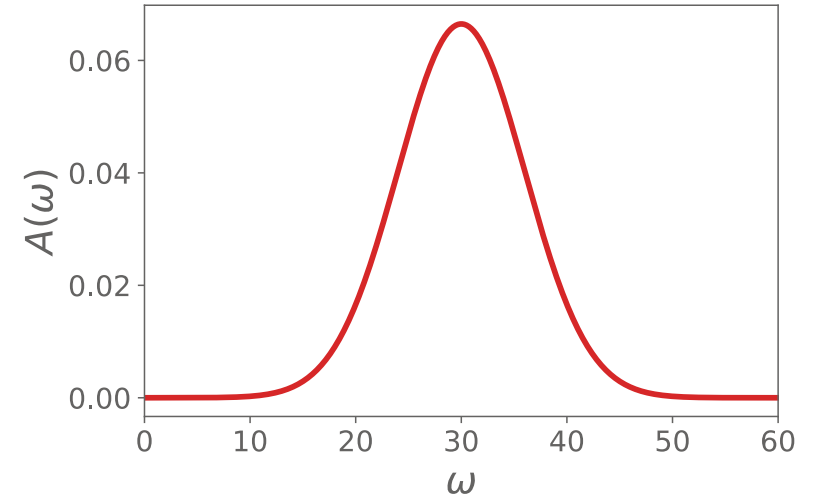
Green function:

$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

# The problem

Green function:

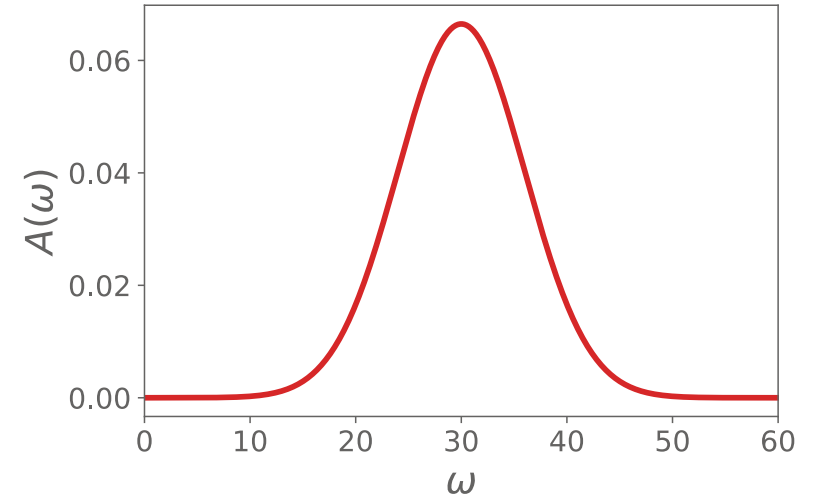
$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$



# The problem

Green function:

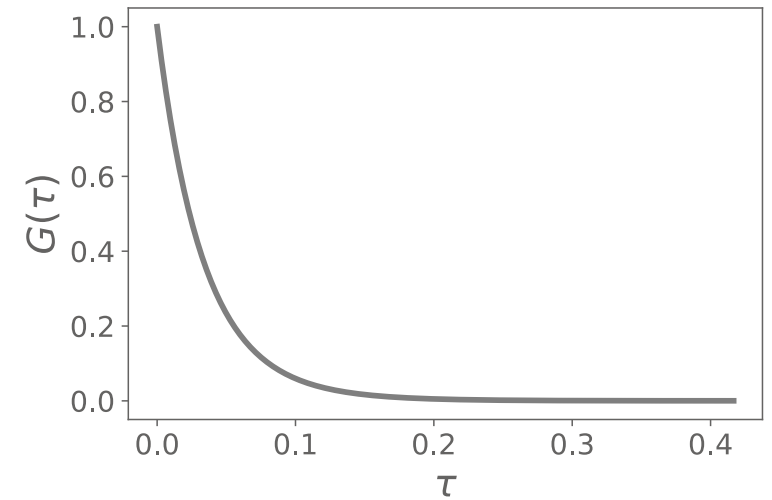
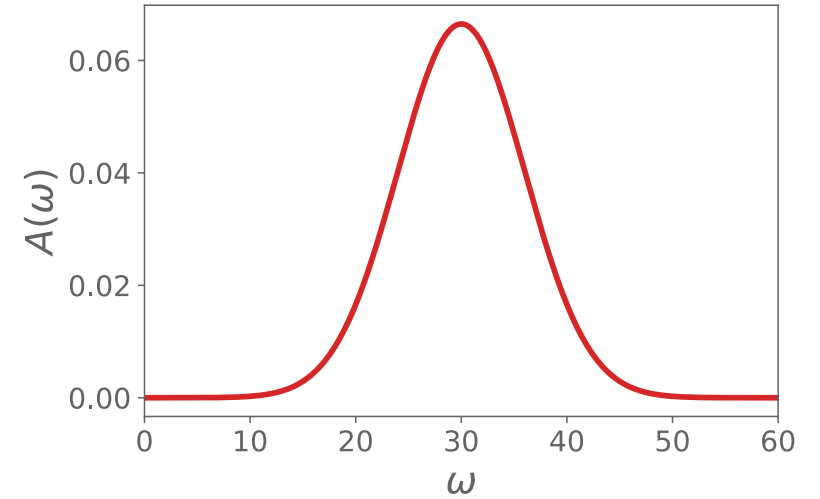
$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$



# The problem

Green function:

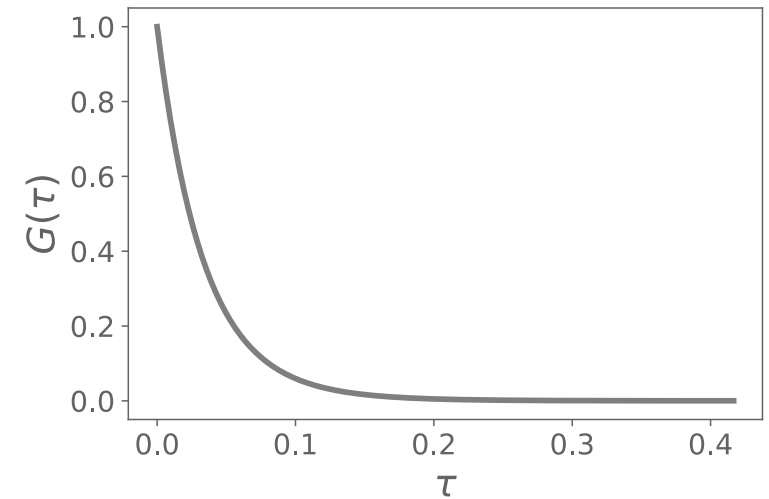
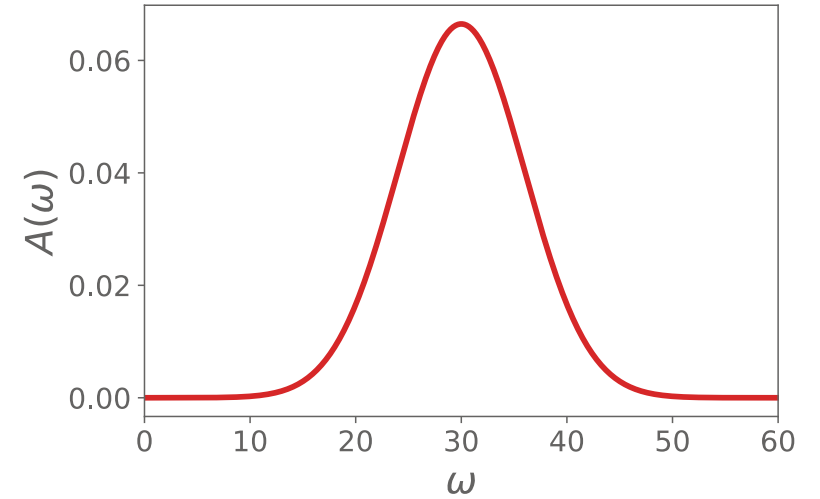
$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$



# The problem

Green function:

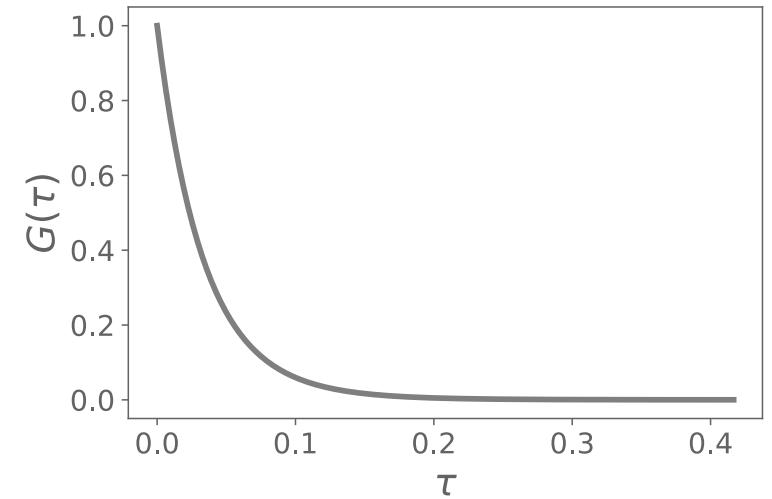
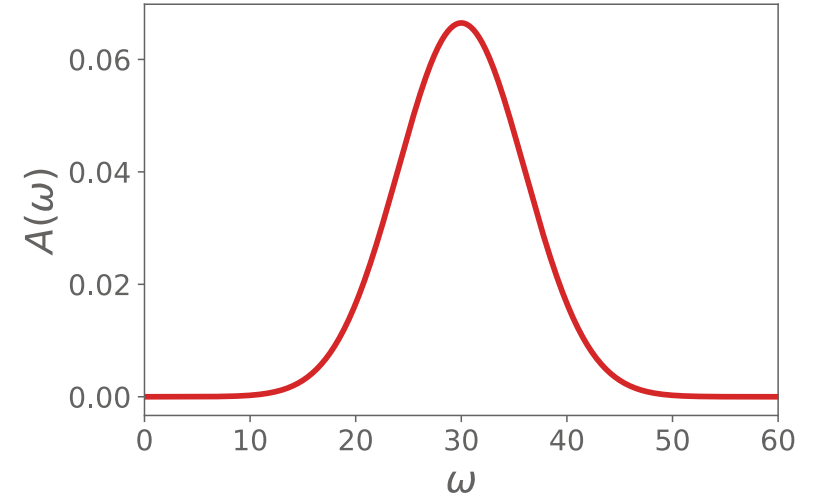
$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$



# The problem

Green function:

$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$



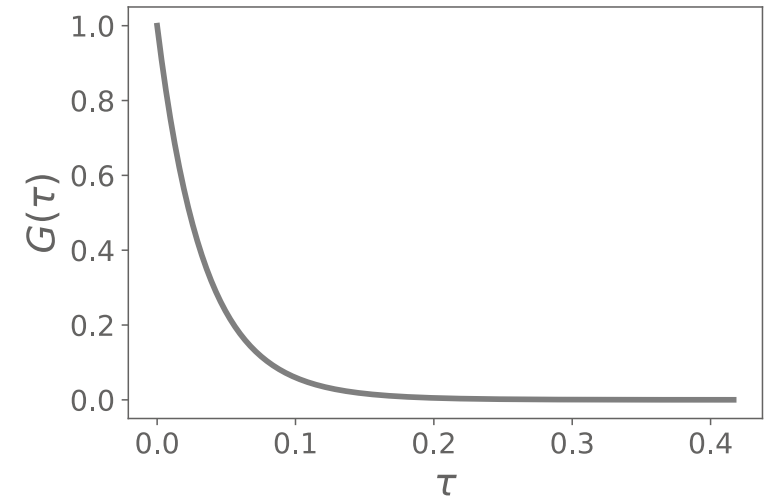
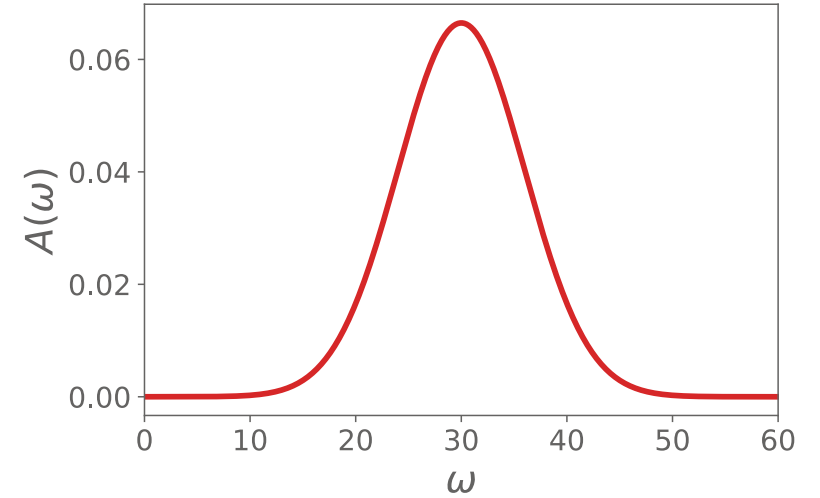
# The problem

Green function:

$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

Spectral function:

$$A(\omega) = ?$$





# The problem

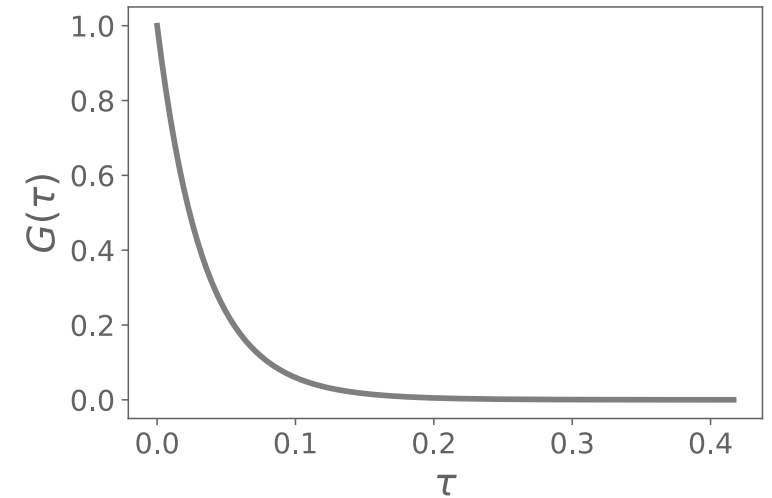
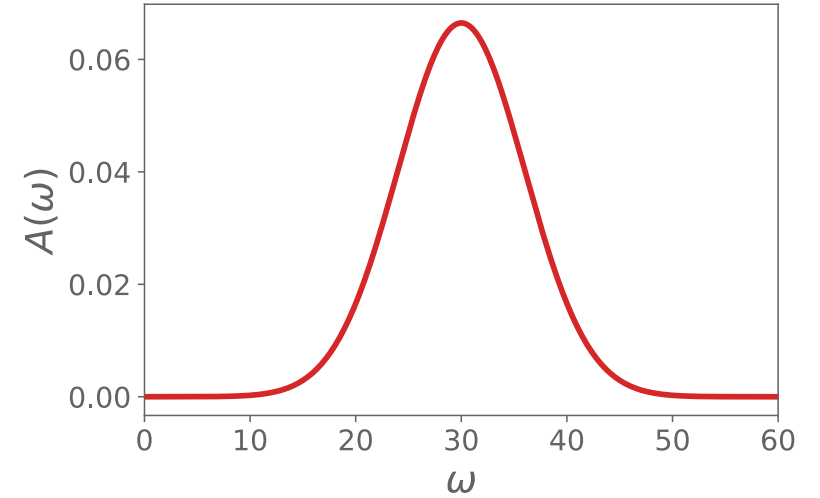
Green function:

$$G(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

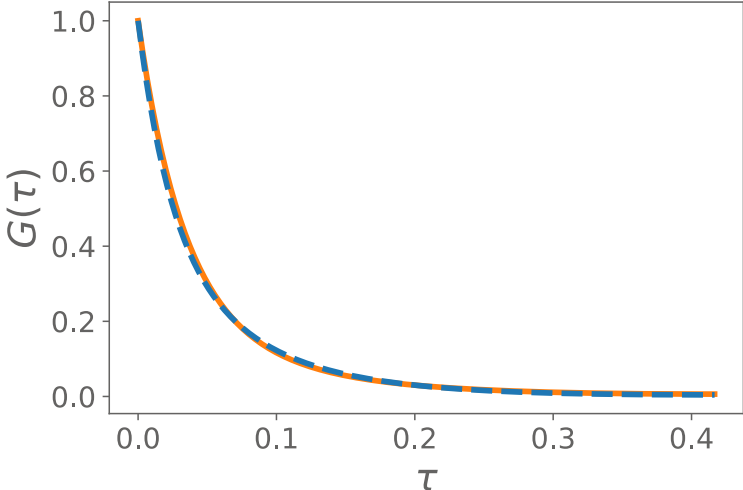
$$G(\tau) = \sum \Delta\omega K(\tau, \omega_i) A(\omega_i)$$

Spectral function:

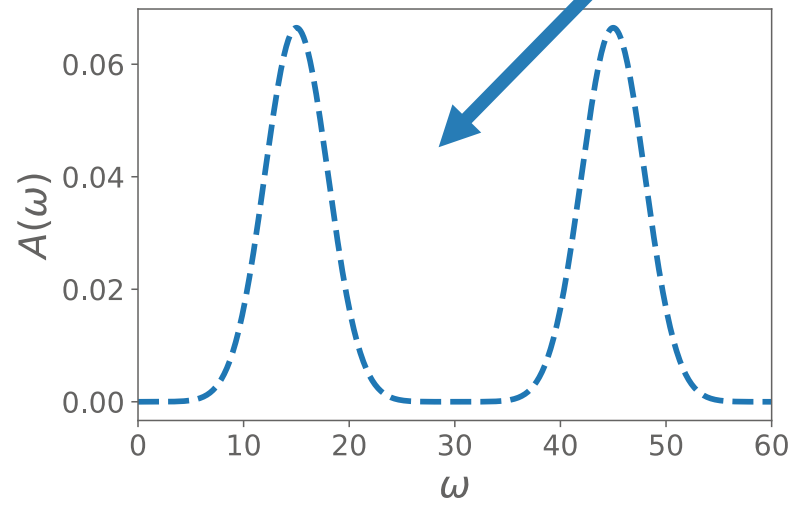
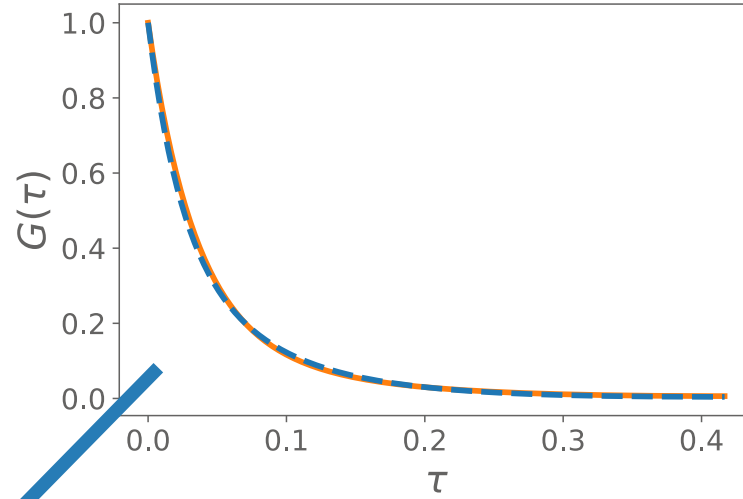
$$A(\omega) = ?$$



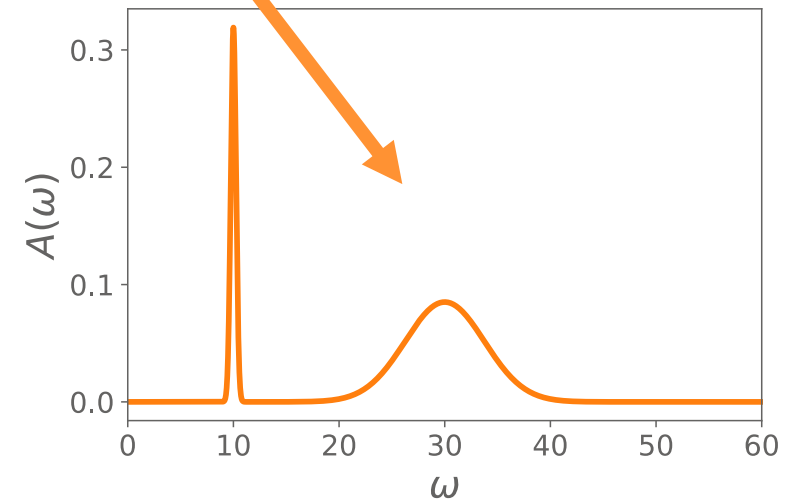
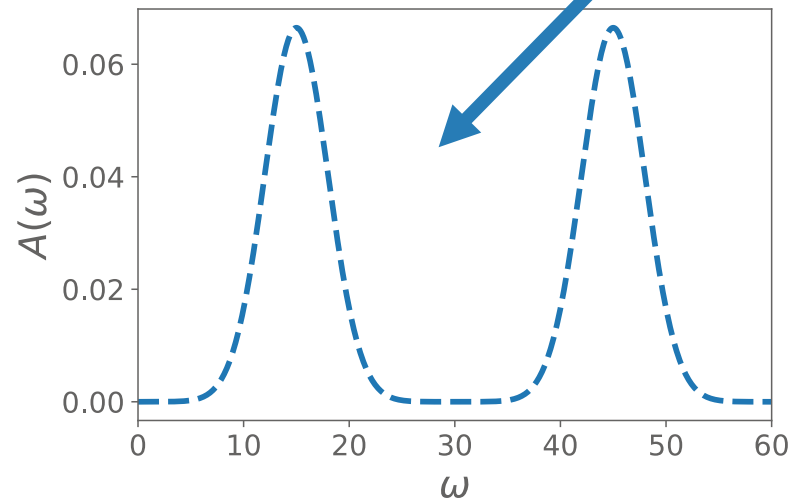
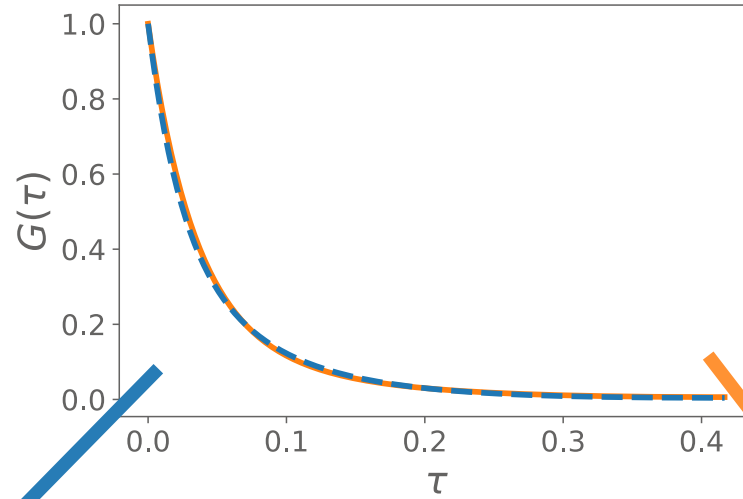
# An example



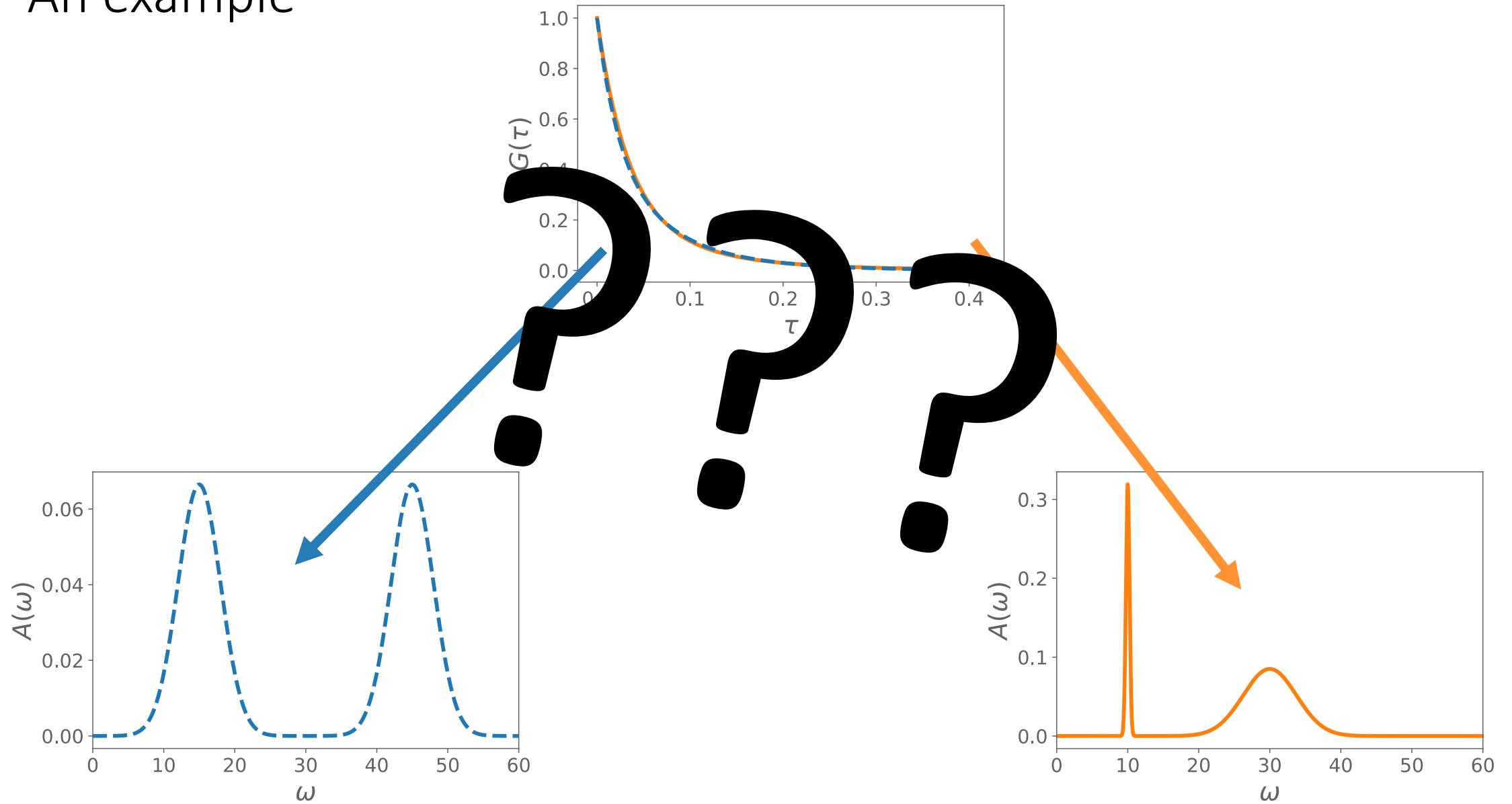
# An example



# An example



# An example



# MaxEnt

Bayes' rule:

$$P(A|G) = \frac{P(G|A)P(A)}{P(G)}$$

# MaxEnt

Bayes' rule:

$$P(A|G) = \frac{P(G|A)P(A)}{P(G)}$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

# MaxEnt

Bayes' rule:

$$P(A|G) = \frac{P(G|A)P(A)}{P(G)}$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

Prior probability:

$$P(A) \propto e^{\alpha S}$$



# MaxEnt

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

# MaxEnt

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

# MaxEnt

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

# MaxEnt

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Prior probability:

$$P(A) \propto e^{\alpha S}$$

# MaxEnt

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Prior probability:

$$P(A) \propto e^{\alpha S}$$

$$S = - \int \frac{d\omega}{2\pi} A(\omega) \ln \frac{A(\omega)}{D(\omega)}$$

# MaxEnt

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Prior probability:

$$P(A) \propto e^{\alpha S}$$

$$S = - \int \frac{d\omega}{2\pi} A(\omega) \ln \frac{A(\omega)}{D(\omega)}$$

$$P(G|A)P(A) \propto e^{\alpha S - \frac{\chi^2}{2}}$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$



# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Algorithm:

$$A_i$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

$$A_{i+\frac{1}{2}} = A_i + \lambda_i$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

$$A_{i+\frac{1}{2}} = A_i + \lambda_i \rightarrow \chi_{i+\frac{1}{2}}^2$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

$$A_{i+\frac{1}{2}} = A_i + \lambda_i \rightarrow \chi_{i+\frac{1}{2}}^2$$

$$\chi_{i+\frac{1}{2}}^2 \leq \chi_i^2$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

$$A_{i+\frac{1}{2}} = A_i + \lambda_i \rightarrow \chi_{i+\frac{1}{2}}^2$$

$$\chi_{i+\frac{1}{2}}^2 \leq \chi_i^2 \rightarrow A_{i+1} = A_{i+\frac{1}{2}}$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

$$A_{i+\frac{1}{2}} = A_i + \lambda_i \rightarrow \chi_{i+\frac{1}{2}}^2$$

$$\chi_{i+\frac{1}{2}}^2 \leq \chi_i^2 \rightarrow A_{i+1} = A_{i+\frac{1}{2}}$$

$$\chi_{i+\frac{1}{2}}^2 > \chi_i^2$$



# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$

Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:

$$A_i \rightarrow \chi_i^2$$

$$A_{i+\frac{1}{2}} = A_i + \lambda_i \rightarrow \chi_{i+\frac{1}{2}}^2$$

$$\chi_{i+\frac{1}{2}}^2 \leq \chi_i^2 \rightarrow A_{i+1} = A_{i+\frac{1}{2}}$$

$$\chi_{i+\frac{1}{2}}^2 > \chi_i^2 \rightarrow A_{i+1} = A_i$$

# Fast and Efficient Stochastic Optimization Method (FESOM)

Minimal knowledge of prior probability:

$$P(A|G) = P(G|A)$$


Likelihood:

$$P(G|A) \propto e^{-\frac{\chi^2}{2}}$$

$$\tilde{G}(\tau) = \int d\omega K(\tau, \omega) A(\omega)$$

$$\chi^2 = \frac{1}{N} \sum_i \frac{(\tilde{G}_i - G_i)^2}{\sigma^2}$$

Algorithm:


$$A_i \rightarrow \chi_i^2$$

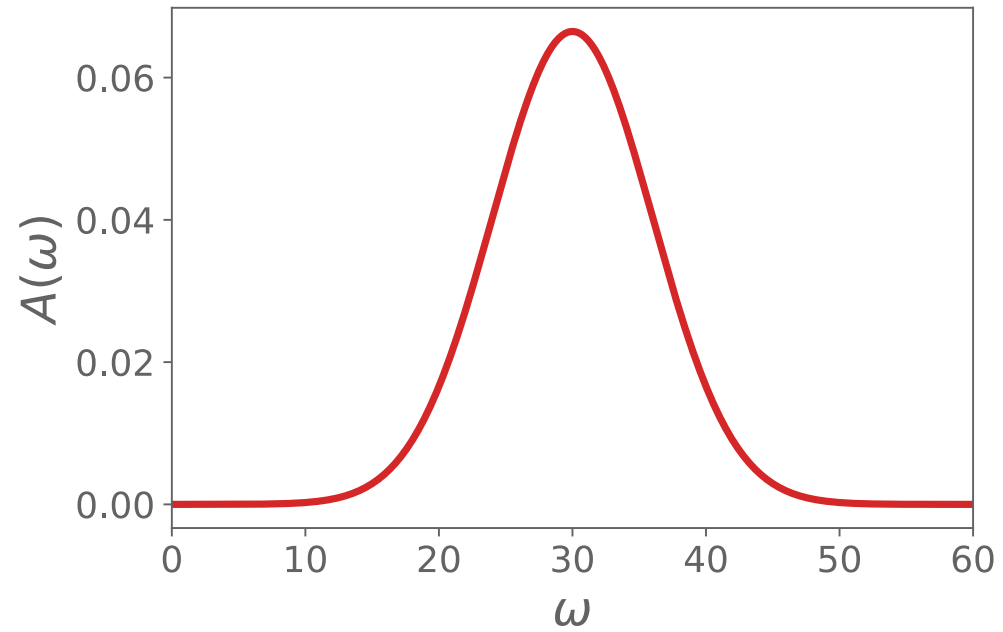
$$A_{i+\frac{1}{2}} = A_i + \lambda_i \rightarrow \chi_{i+\frac{1}{2}}^2$$

$$\chi_{i+\frac{1}{2}}^2 \leq \chi_i^2 \rightarrow A_{i+1} = A_{i+\frac{1}{2}}$$

$$\chi_{i+\frac{1}{2}}^2 > \chi_i^2 \rightarrow A_{i+1} = A_i$$

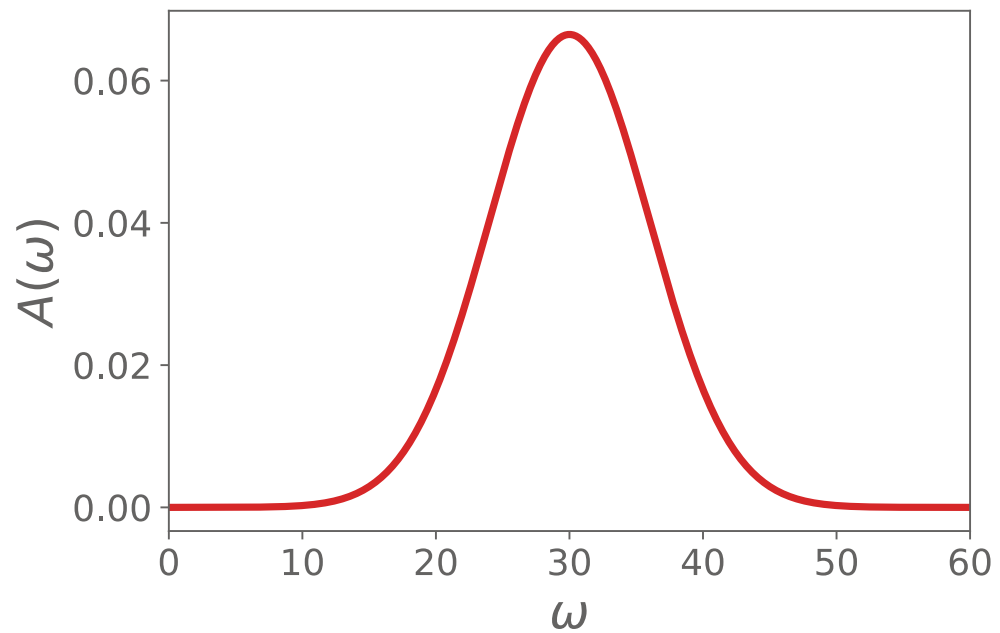
# FESOM example

$A_i$

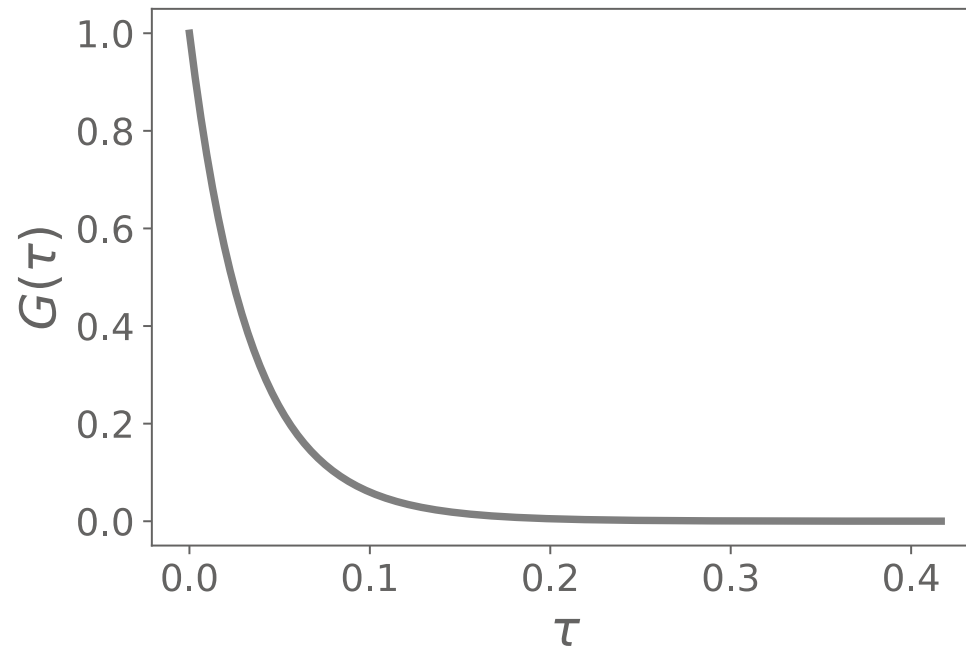


# FESOM example

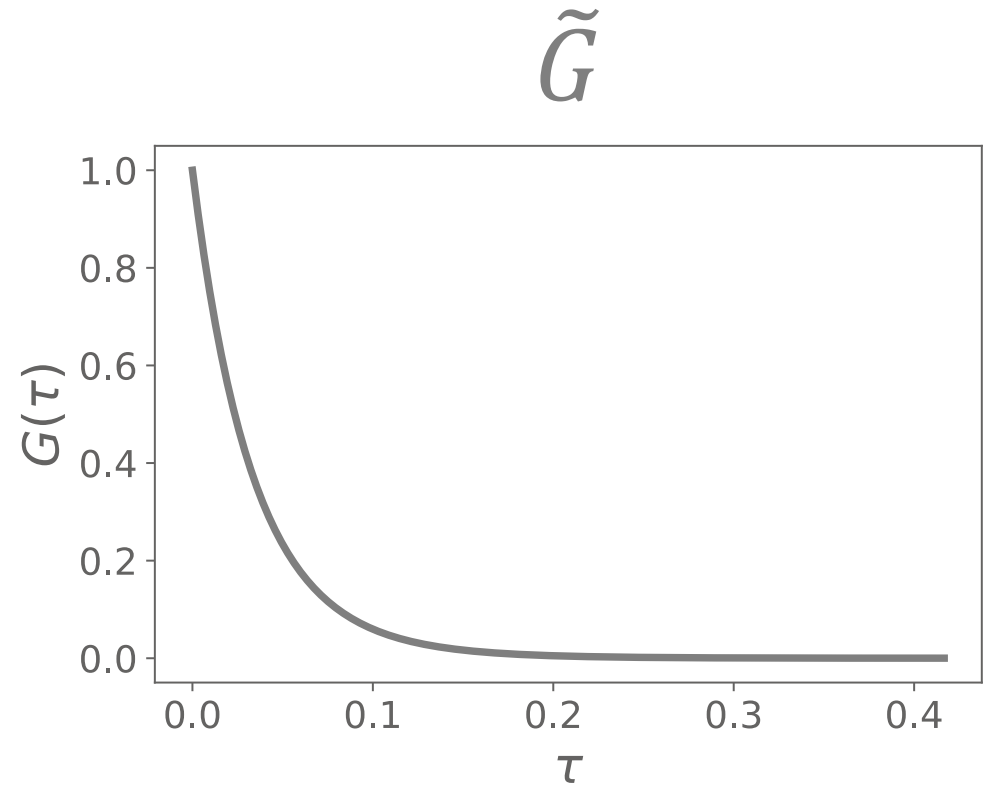
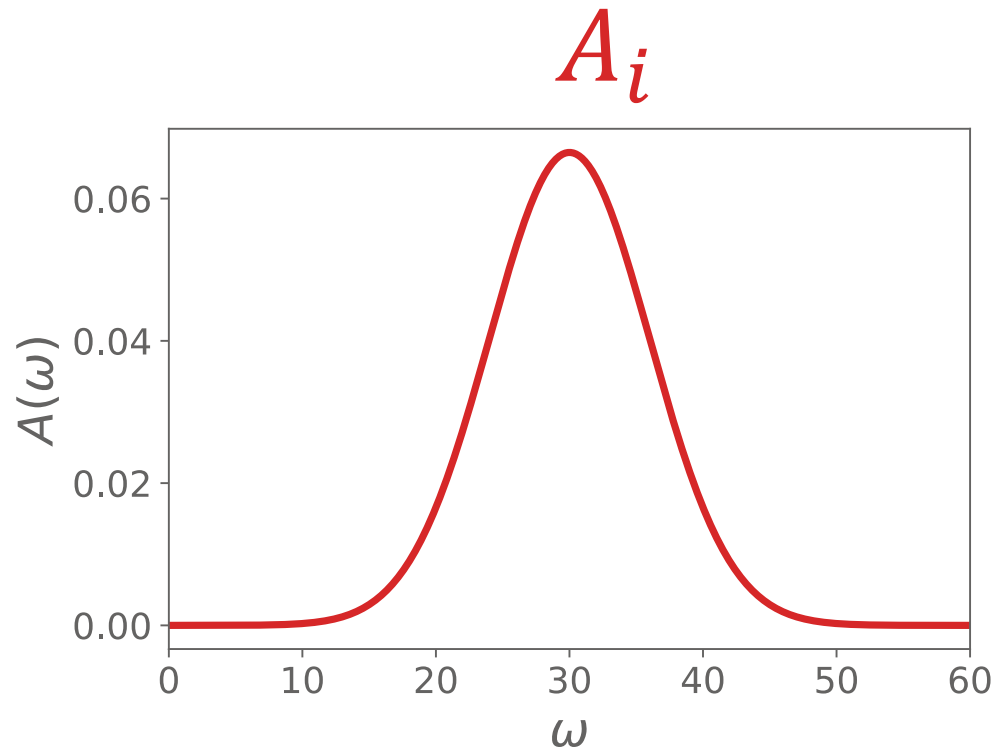
$A_i$



$\tilde{G}$



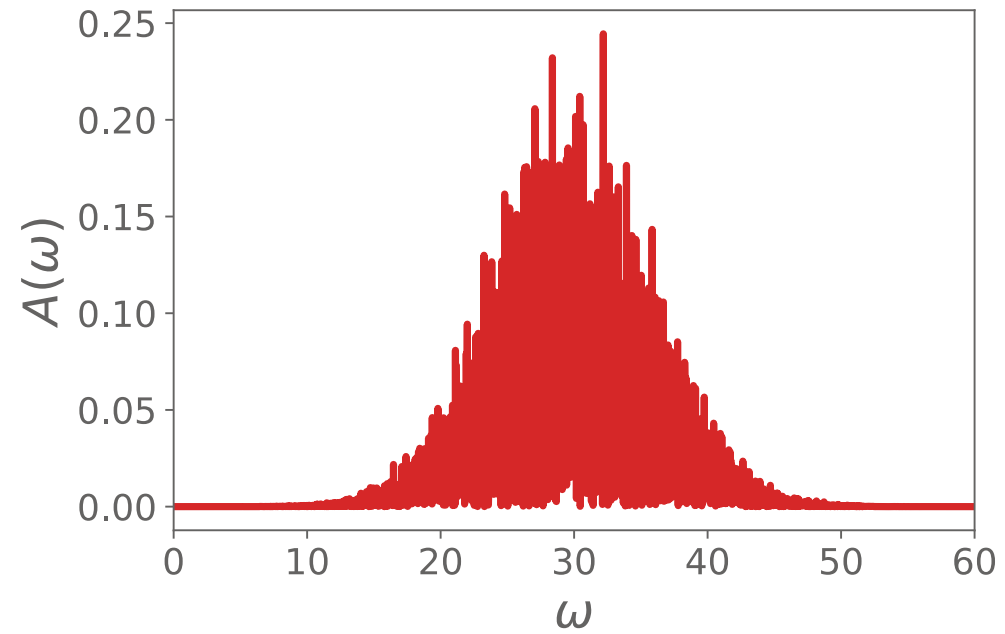
# FESOM example



$$\chi_i^2 = 5.53$$

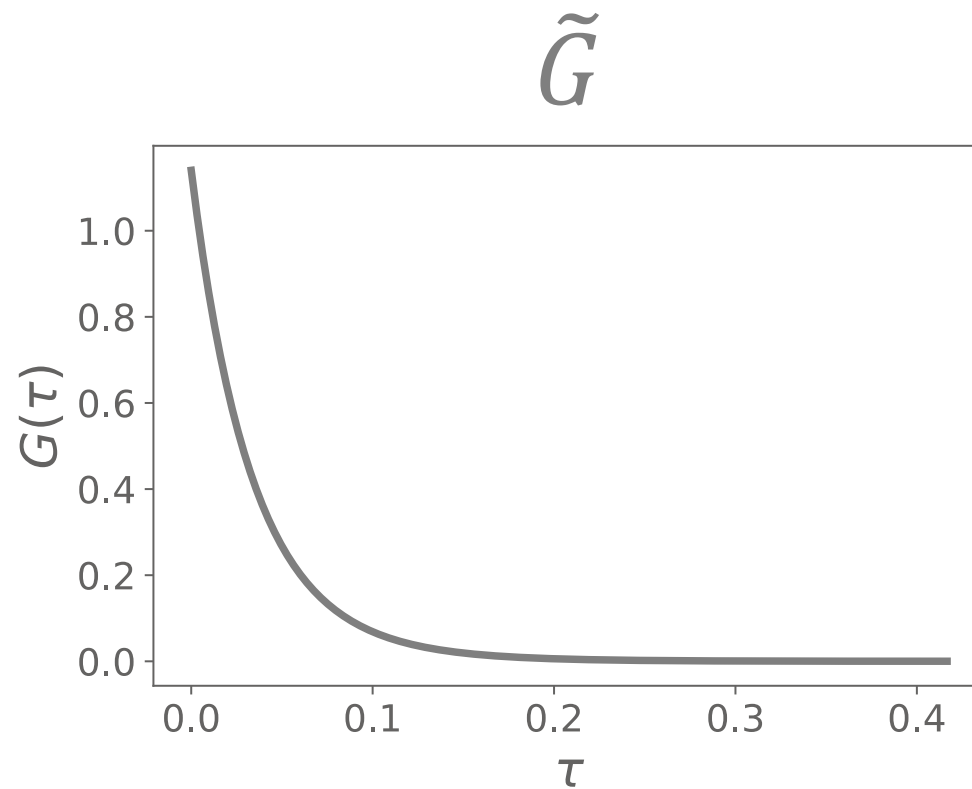
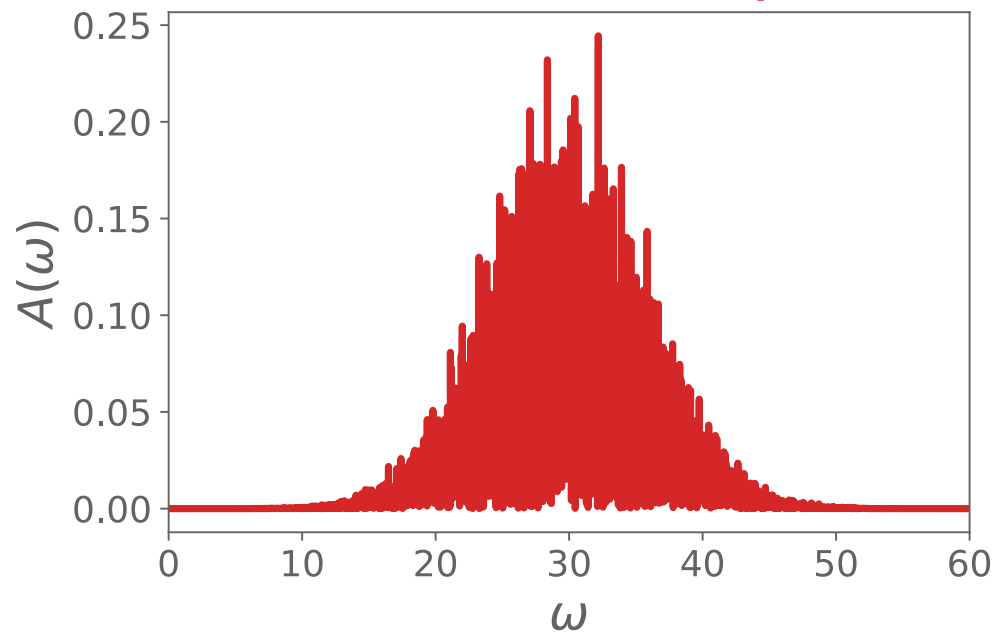
# FESOM example

$A_{i+1/2}$

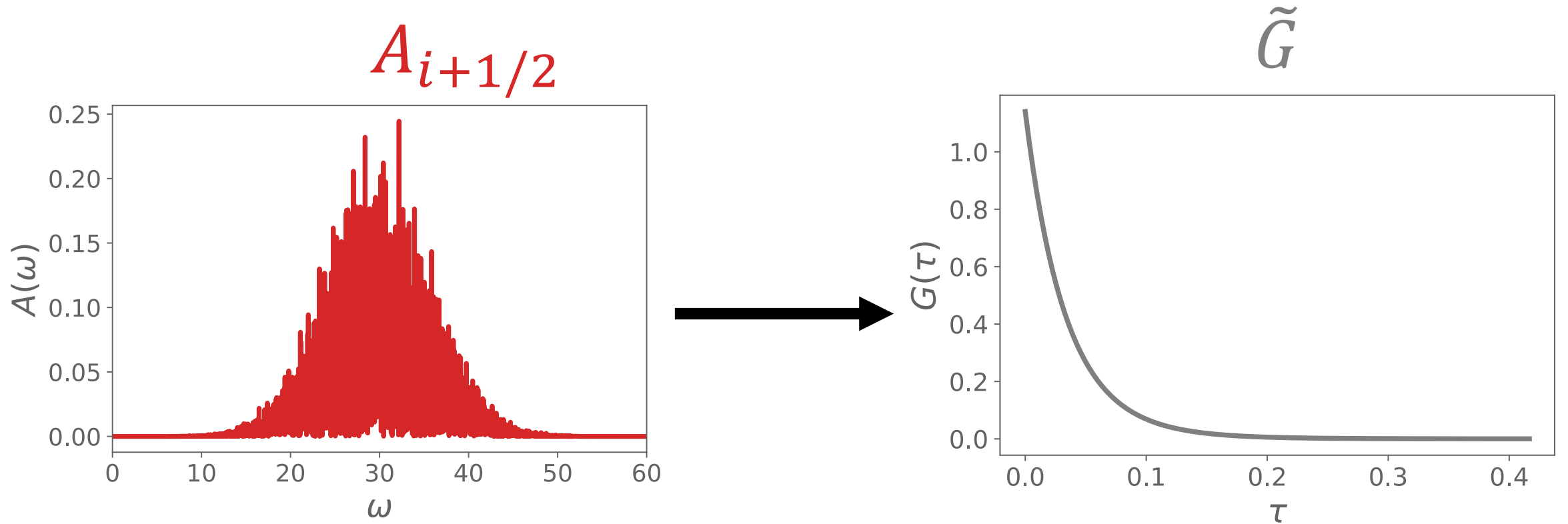


# FESOM example

$A_{i+1/2}$



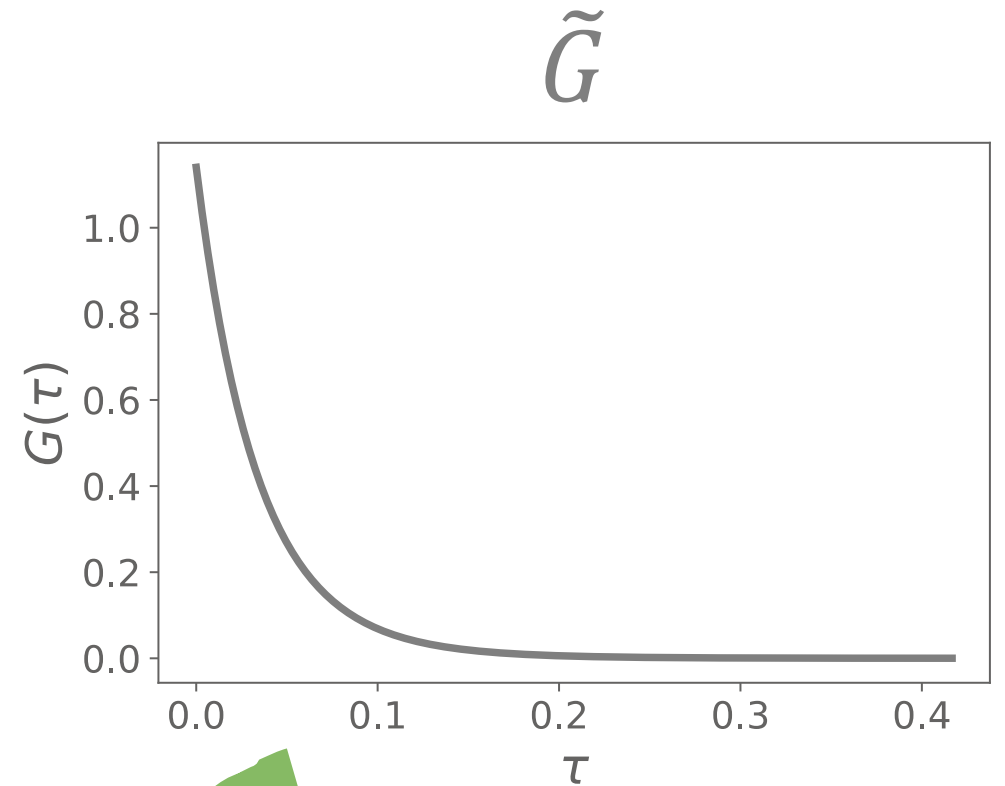
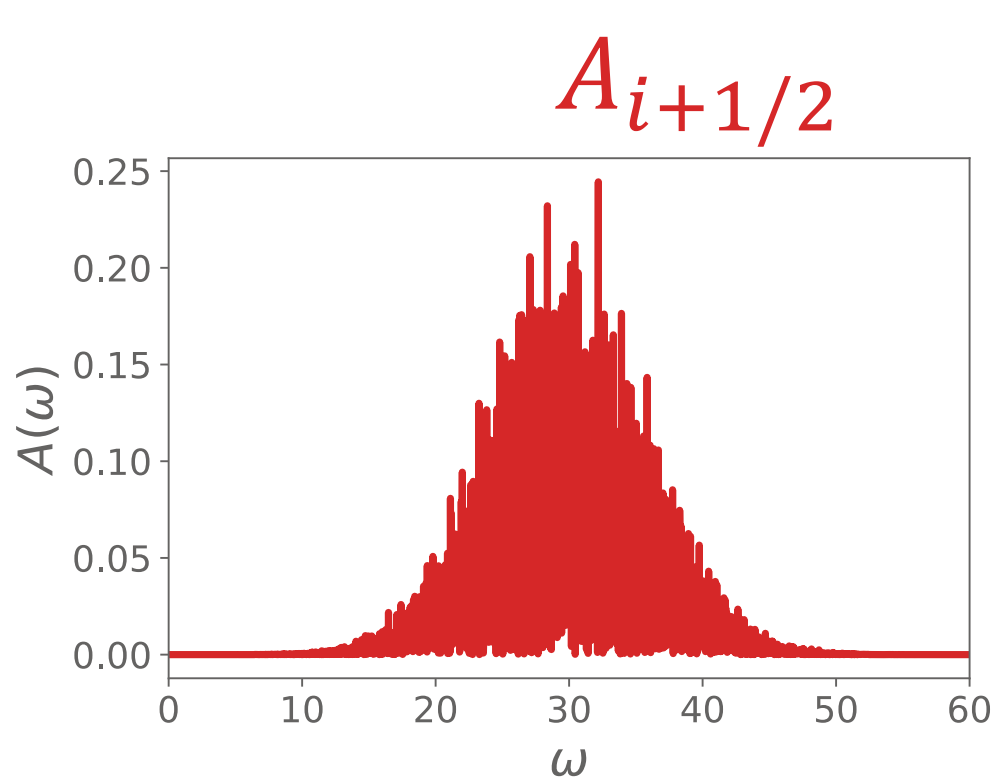
# FESOM example



$$\chi_{i+1/2}^2 = 5.25$$



# FESOM example

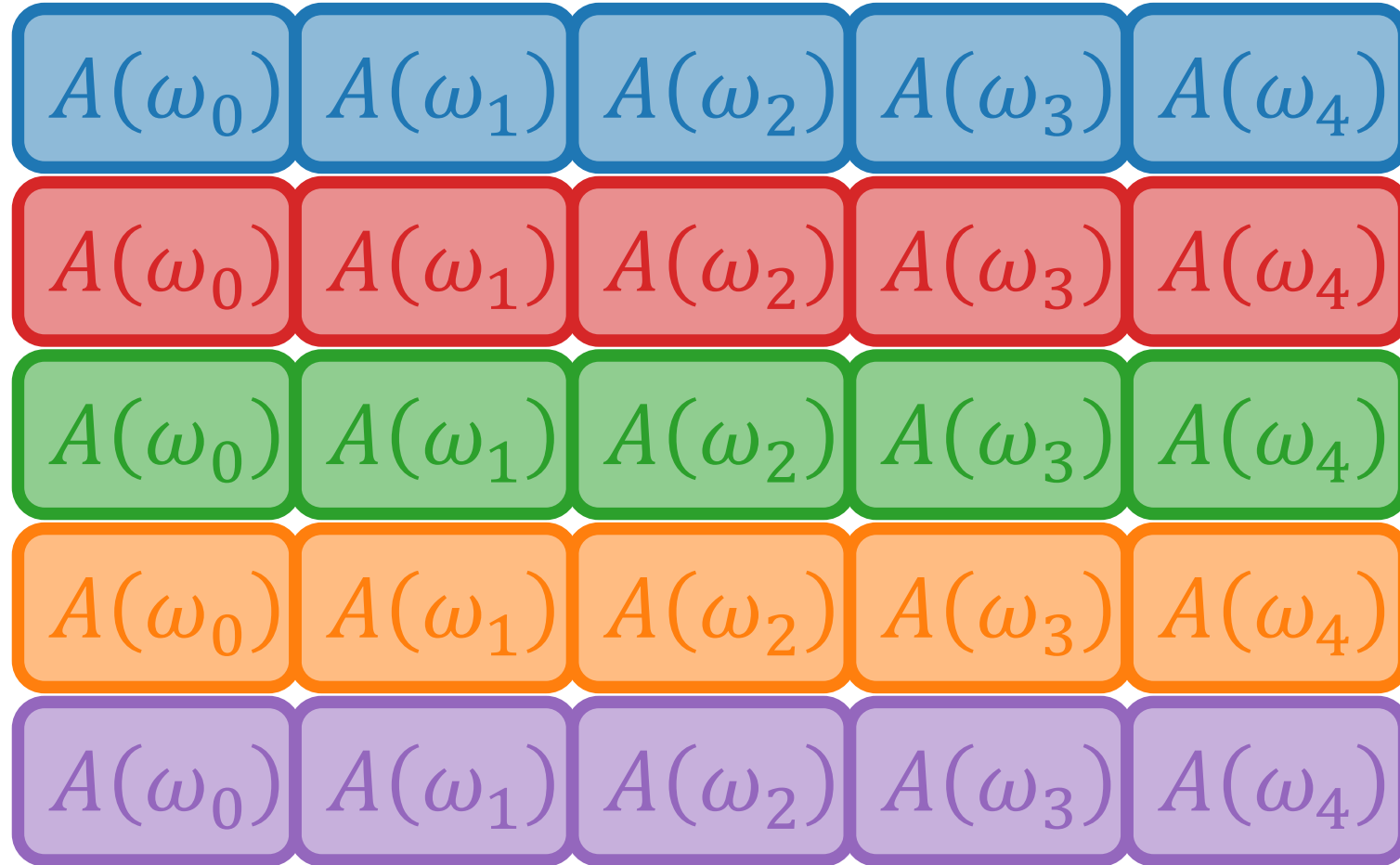


$$\chi_{i+1/2}^2 = 5.25$$

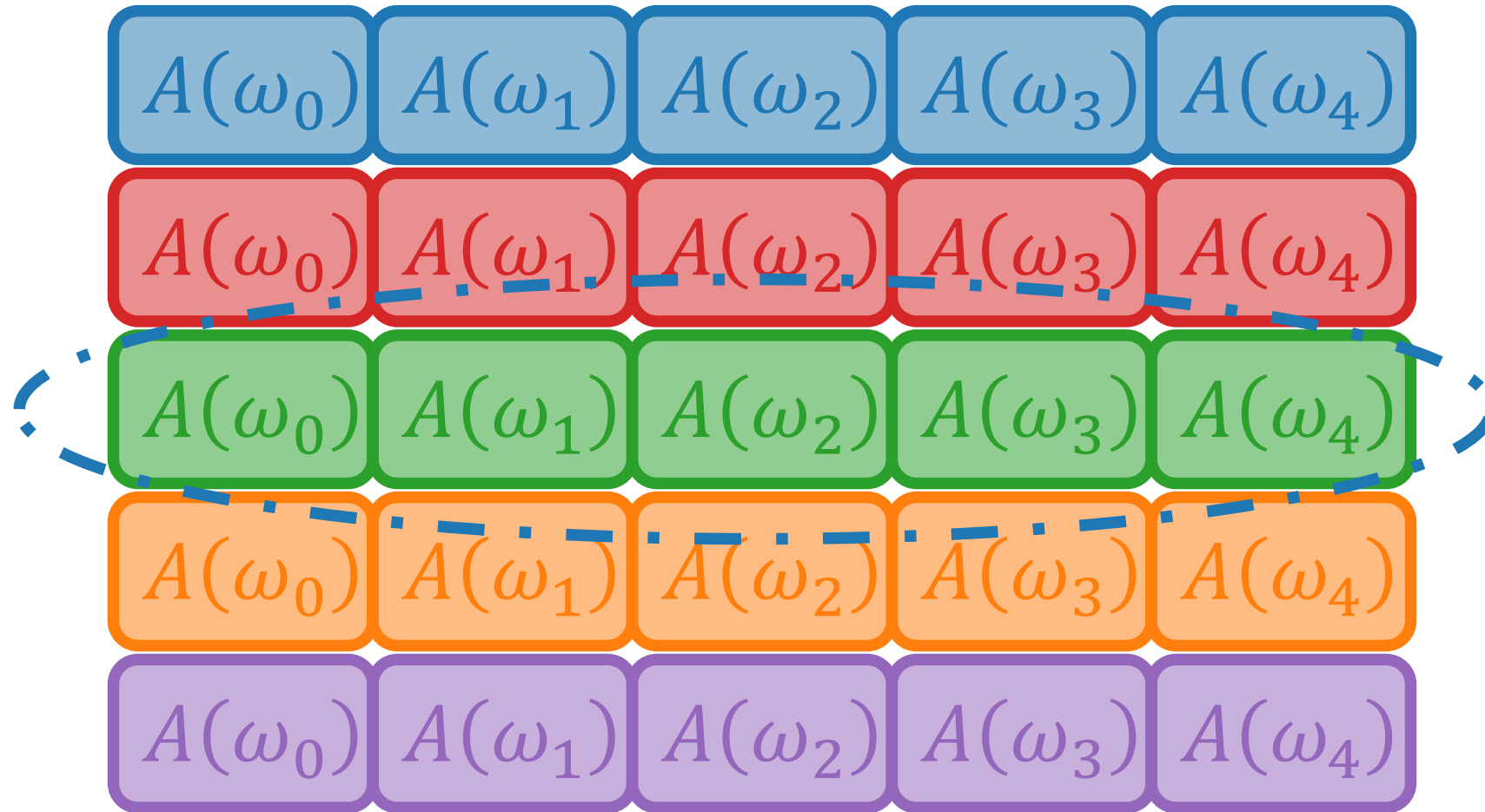
# Differential Evolution for Analytic Continuation (DEAC)

- Inspired by GIFT
- Evolutionary algorithm
  - Population of individuals
  - Genome  $\rightarrow A(\omega_i)$
  - Mutations  $\rightarrow$  vector differences
  - Fitness  $\rightarrow f(A; \chi^2, \langle \omega^k \rangle)$
  - Rejection  $\rightarrow$  keep most fit

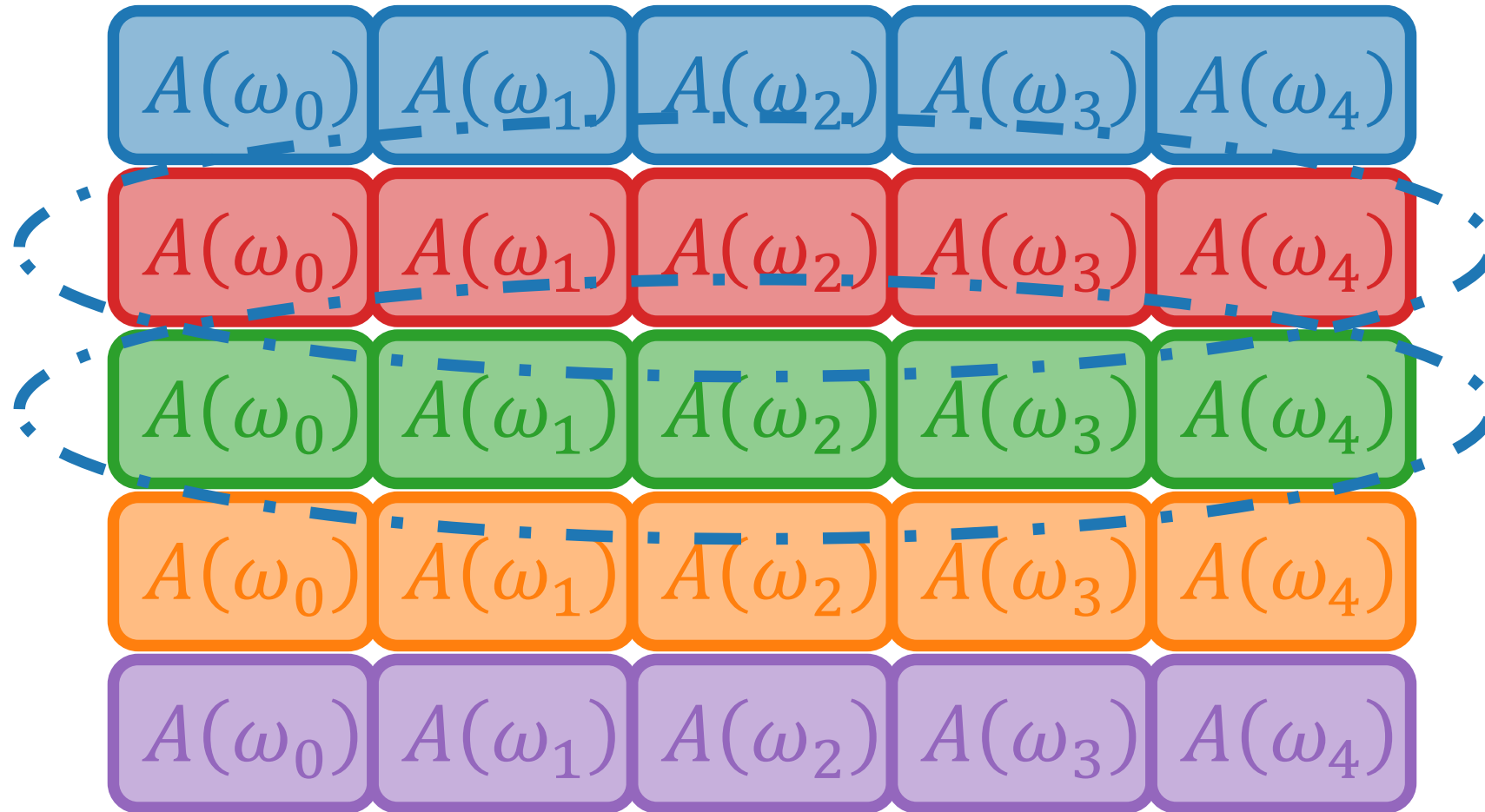
# DEAC Population



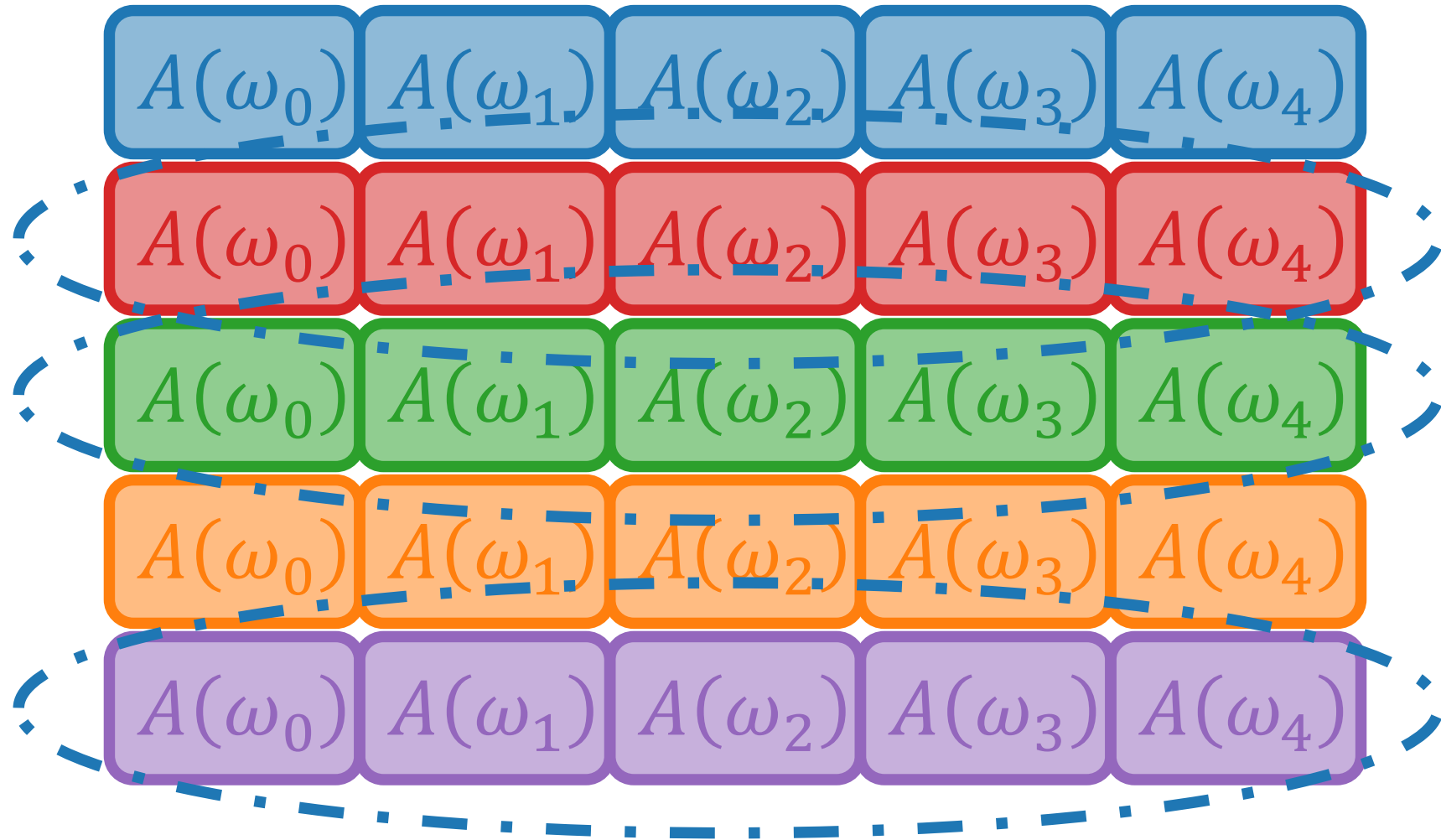
# DEAC Mutation



# DEAC Mutation



# DEAC Mutation



# DEAC Mutation

$$\tilde{A}(\omega_i)$$

# DEAC Mutation

$$U(0,1) < C$$

$$\tilde{A}(\omega_i)$$



# DEAC Mutation

$$U(0,1) < C$$

$$\tilde{A}(\omega_i) = A(\omega_i) + F \times (A(\omega_i) - A(\omega_i))$$

# DEAC Mutation

$$U(0,1) < C$$

$$\tilde{A}(\omega_i) = A(\omega_i) + F \times (A(\omega_i) - A(\omega_i))$$

$$U(0,1) \geq C$$

$$\tilde{A}(\omega_i) = A(\omega_i)$$

# DEAC Mutation

$$U(0,1) < C$$

$$\tilde{A}(\omega_i) = A(\omega_i) + F \times (A(\omega_i) - A(\omega_i))$$

$$U(0,1) \geq C$$

$$\tilde{A}(\omega_i) = A(\omega_i)$$

$$f(\tilde{A}) < f(A)$$

# DEAC Mutation

$$U(0,1) < C$$

$$\tilde{A}(\omega_i) = A(\omega_i) + F \times (A(\omega_i) - A(\omega_i))$$

$$U(0,1) \geq C$$

$$\tilde{A}(\omega_i) = A(\omega_i)$$

$$f(\tilde{A}) < f(A)$$

# DEAC Mutation

$$U(0,1) < C$$

$$\tilde{A}(\omega_i) = A(\omega_i) + F \times (A(\omega_i) - A(\omega_i))$$

$$U(0,1) \geq C$$

$$\tilde{A}(\omega_i) = A(\omega_i)$$

$f(\tilde{A}) < f(A)$

# DEAC Mutation

$$U(0,1) < C$$

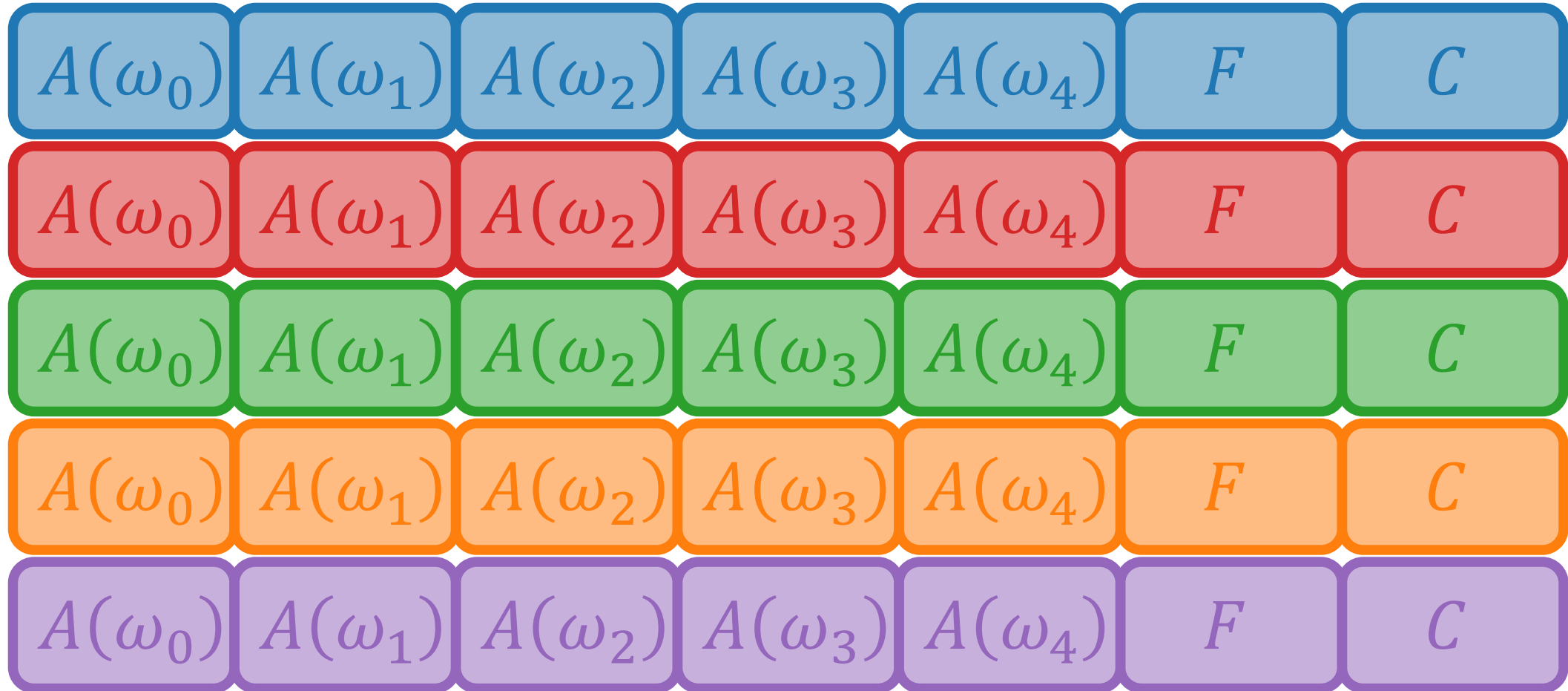
$$\tilde{A}(\omega_i) = A(\omega_i) + F \times (A(\omega_i) - A(\omega_i))$$

$$U(0,1) \geq C$$

$$\tilde{A}(\omega_i) = A(\omega_i)$$

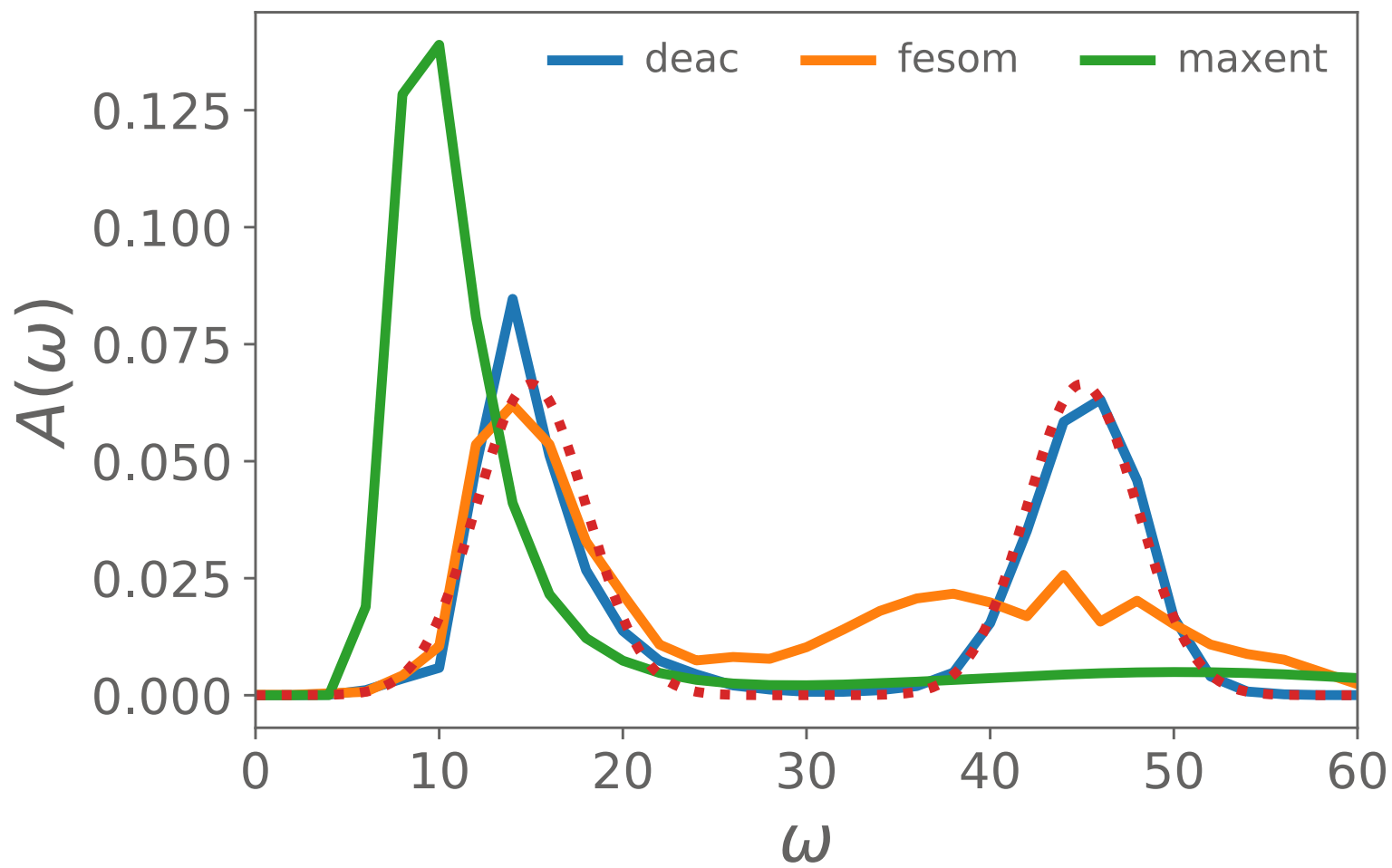
$$f(\tilde{A}) < f(A)$$

# DEAC Population

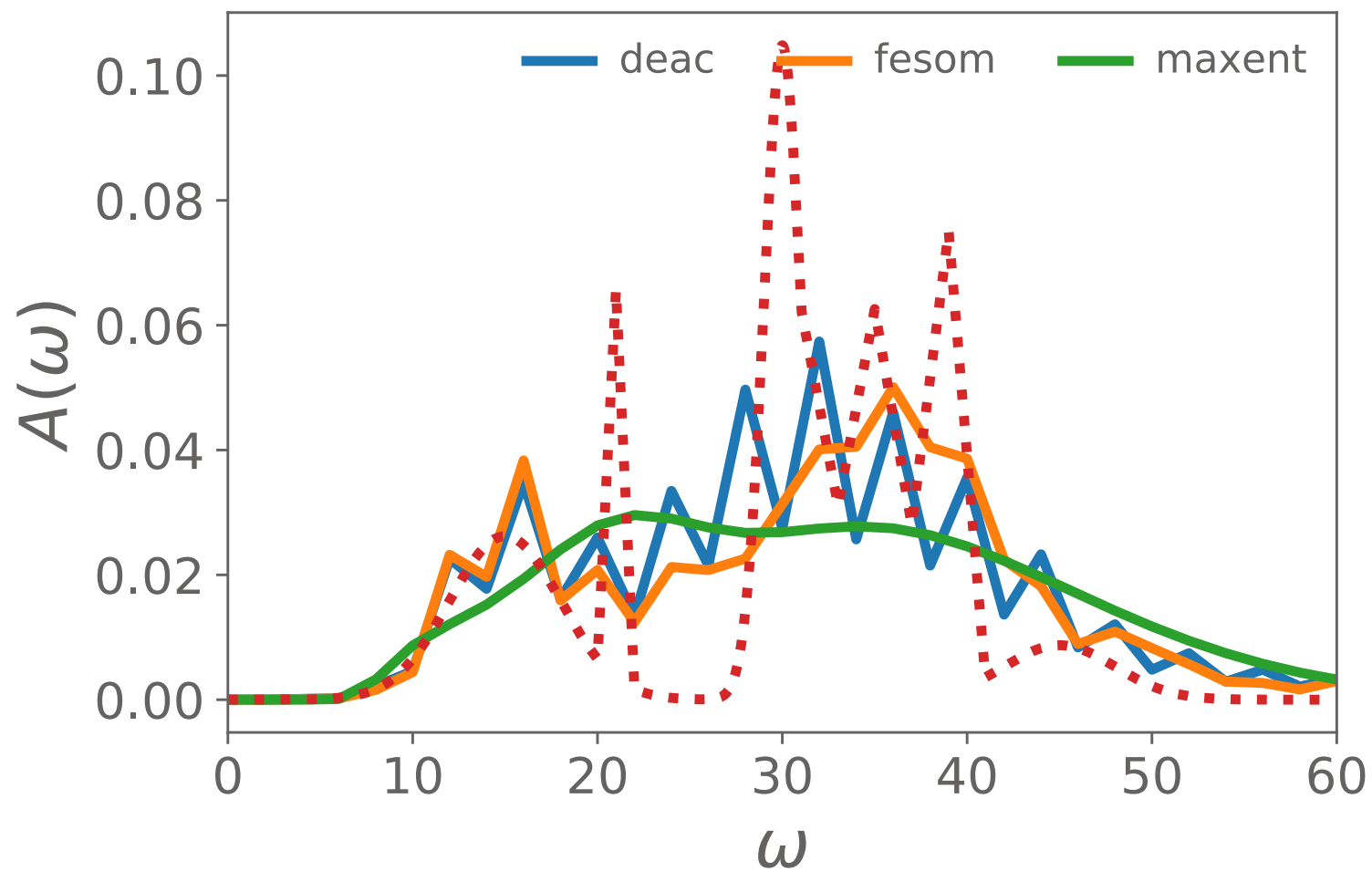


How do the three methods compare?

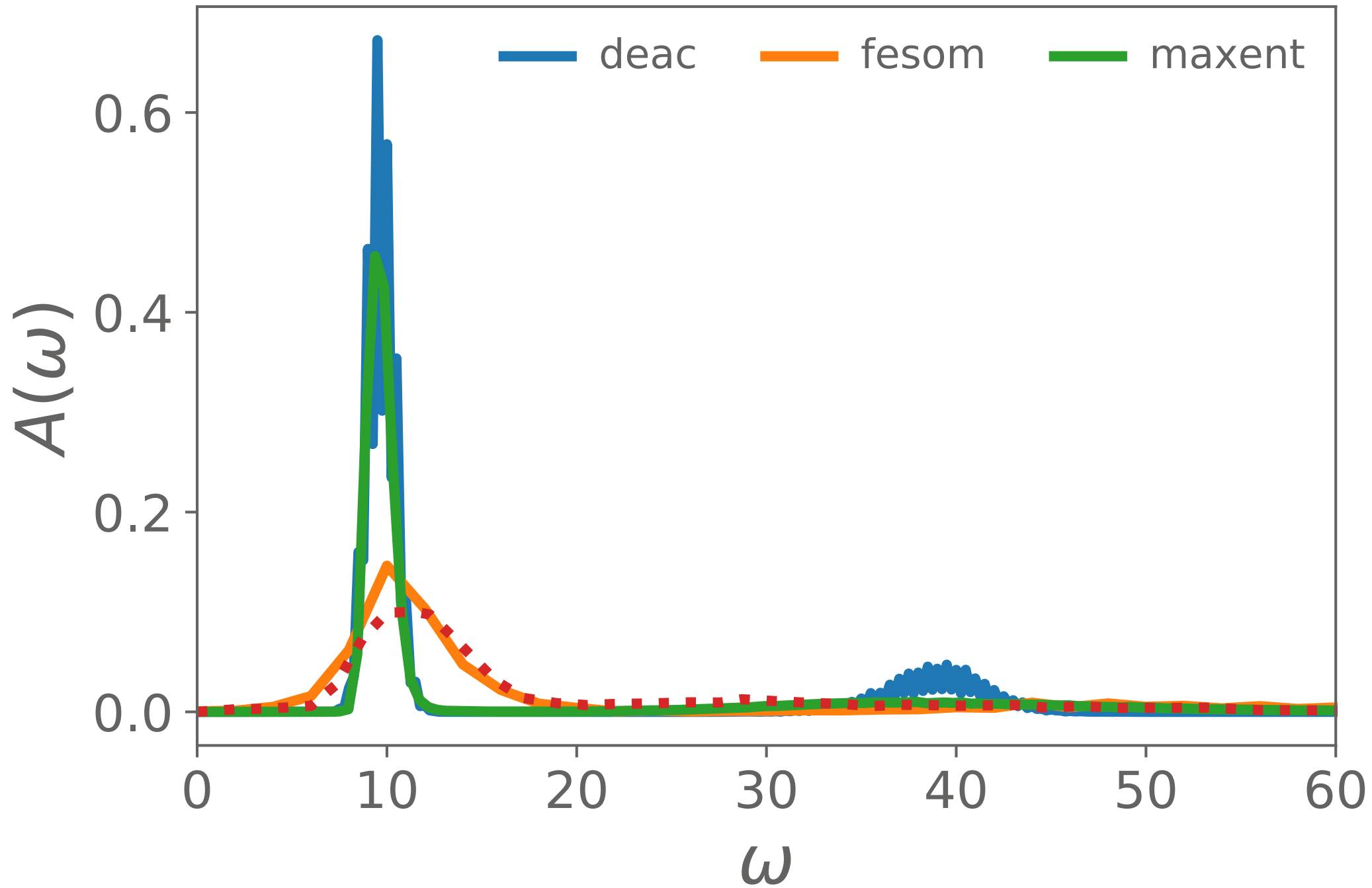




| Method | $\chi^2$             |
|--------|----------------------|
| MEM    | $3.5 \times 10^{-4}$ |
| FESOM  | $5.2 \times 10^{-5}$ |
| DEAC   | $8.6 \times 10^{-6}$ |



| Method | $\chi^2$             |
|--------|----------------------|
| MEM    | $9.3 \times 10^{-5}$ |
| FESOM  | $6.0 \times 10^{-5}$ |
| DEAC   | $8.9 \times 10^{-5}$ |



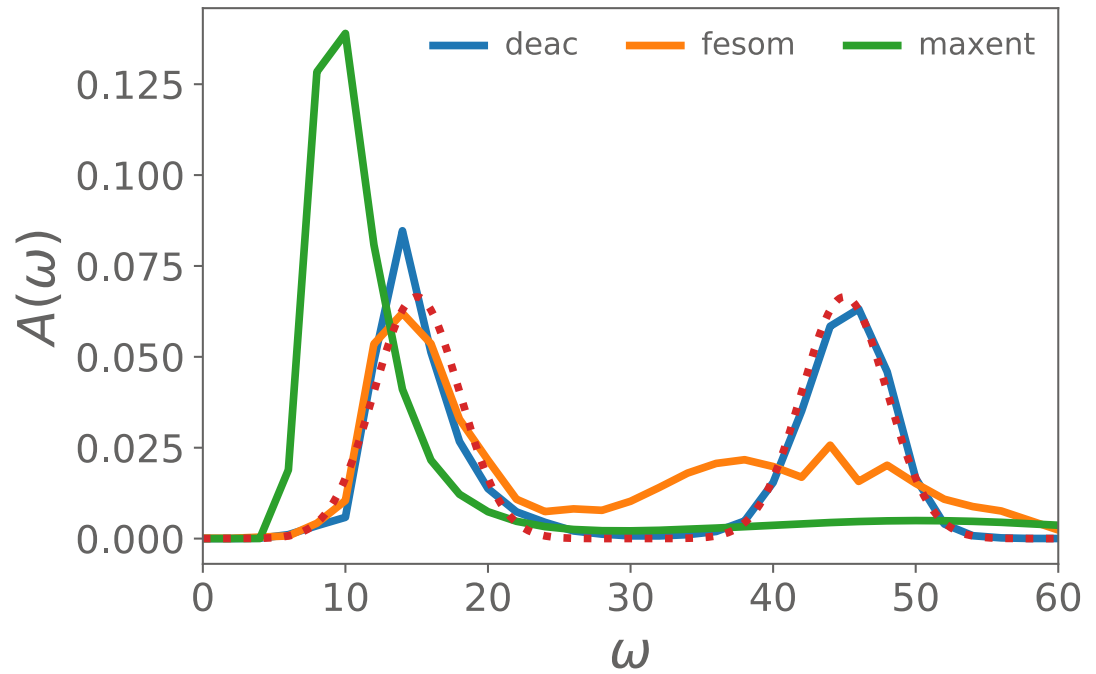
# Future Work

- Use DEAC on new qmc data
- Port code to gpu

**GitHub**

DelMaestroGroup

nscottnichols



The University of Vermont