

# Tensor Networks and spectral properties: probing ETH

Mari-Carmen Bañuls

Maxine Luo, Yilun Yang, Rahul Trivedi, Siri Lu,  
J. Ignacio Cirac

PRB 109, 134304 (2024)

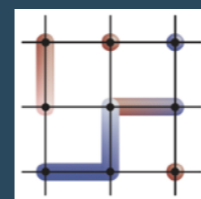
PRB 106, 024307 (2022)

PRX Quantum 2, 020321 (2021)

PRL 124, 100602 (2020)



MAX PLANCK INSTITUTE  
OF QUANTUM OPTICS



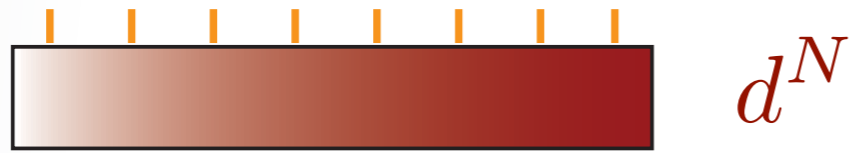
DFG FOR 5522



DFG TRR 360

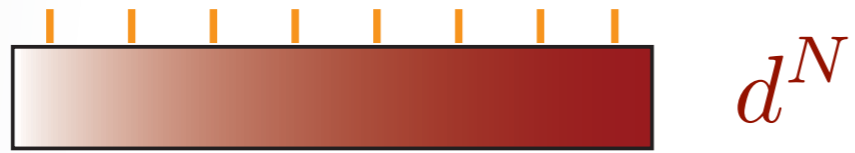
IHP 17.10.2024

arbitrary many-  
body state



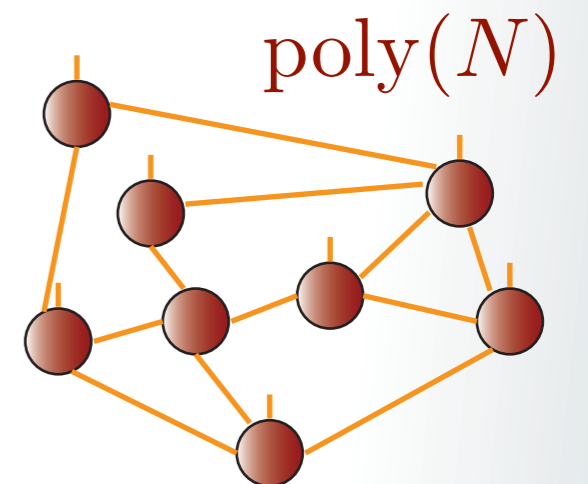
$$|\Psi\rangle = \sum_{i_j} c_{i_1 \dots i_N} |i_1 \dots i_N\rangle$$

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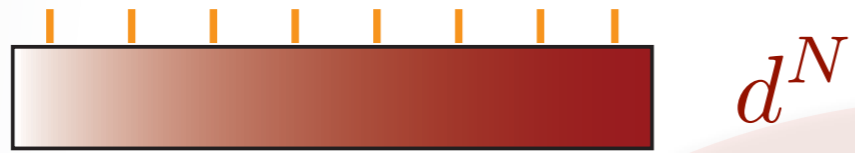


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TNS: restricted  
family

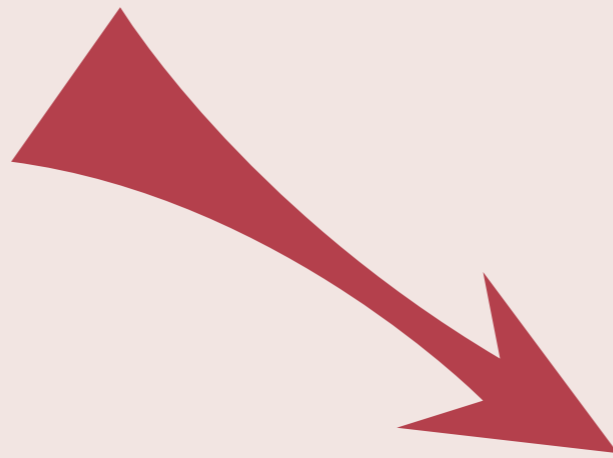


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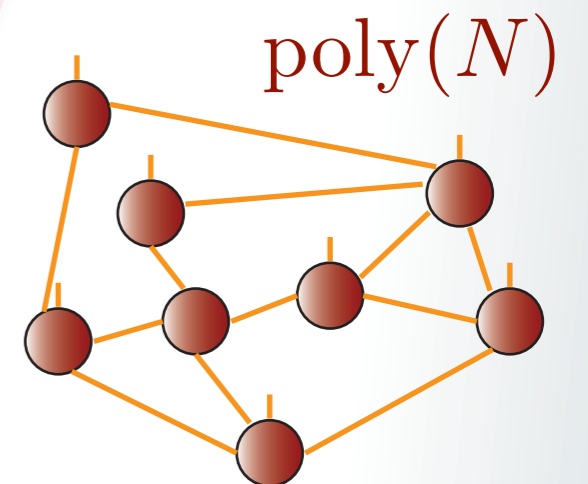
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**exponential**



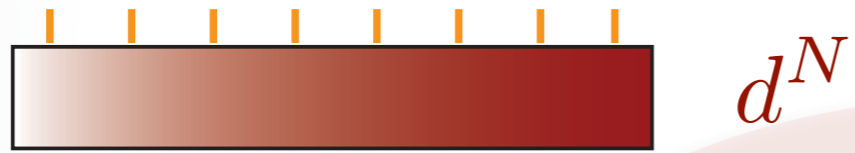
**polynomial**

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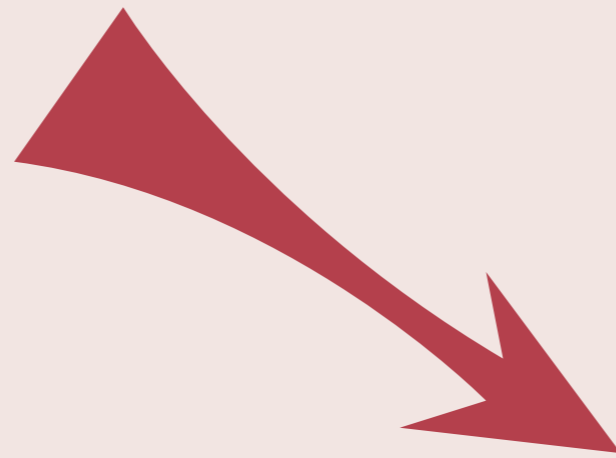


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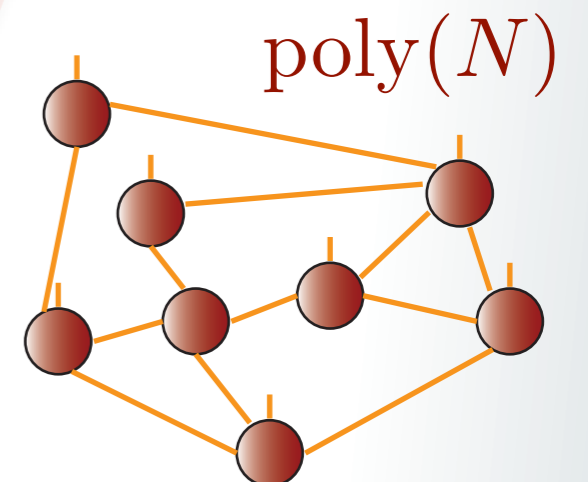
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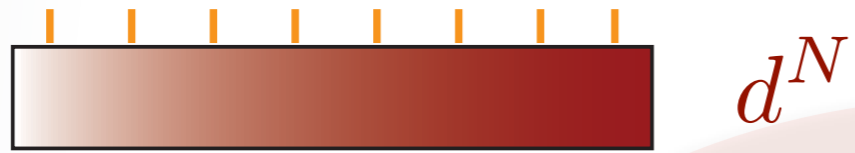
good ansatz for ground states  
and thermal equilibrium: area  
law

entanglement hierarchy

efficient numerics

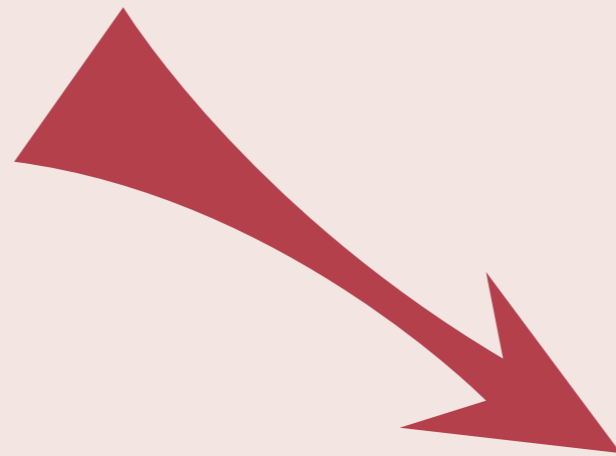


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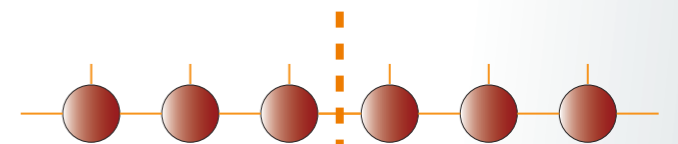
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MPS

matrix product states



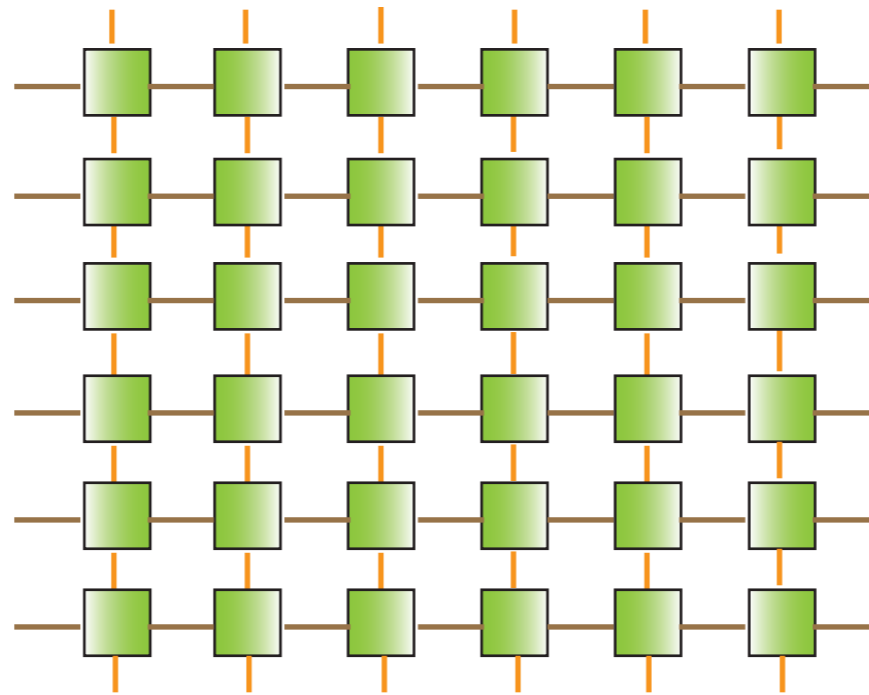
$$S \leq \log(D)$$

TNS: entanglement-based ansatzes for quantum  
many-body states

a side comment

tensor networks may also describe  
partition functions (observables)

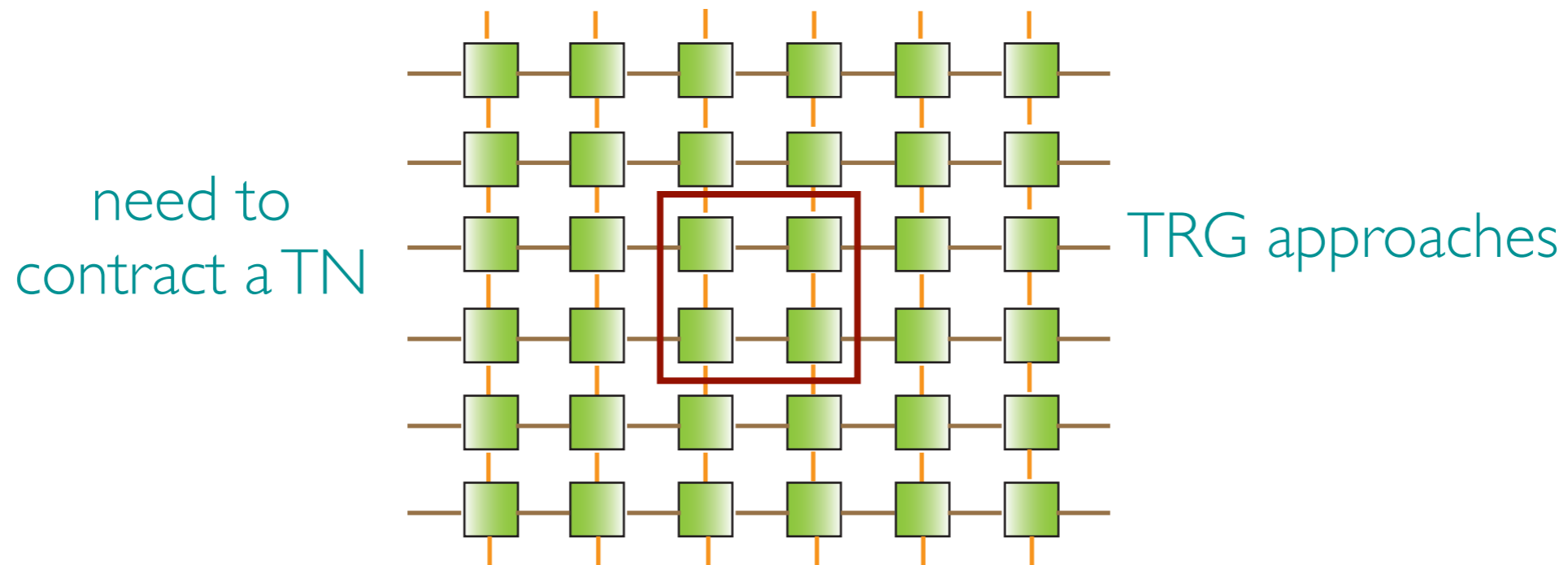
need to  
contract a TN



Nishino, JPSJ 1995  
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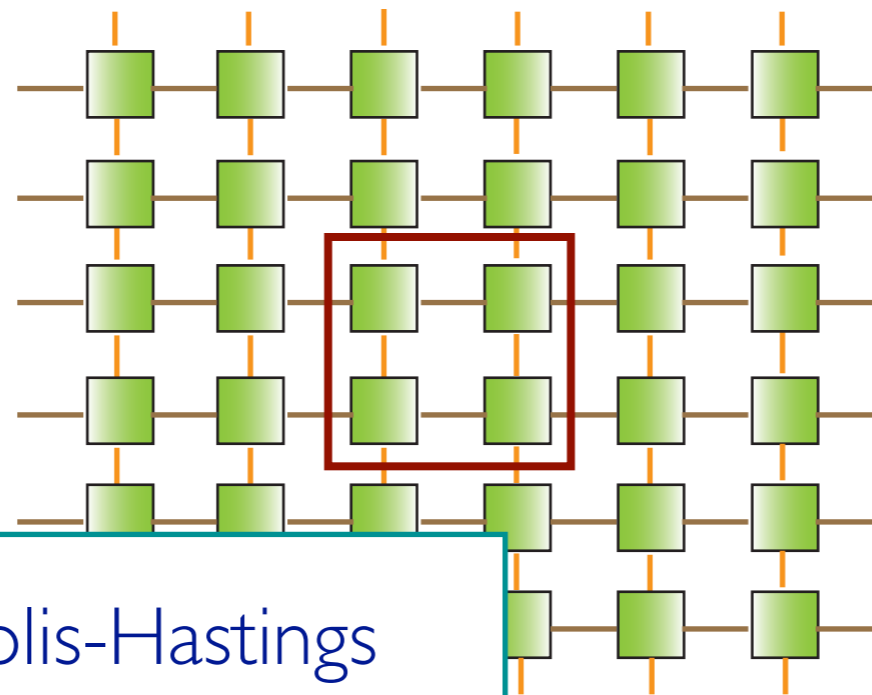
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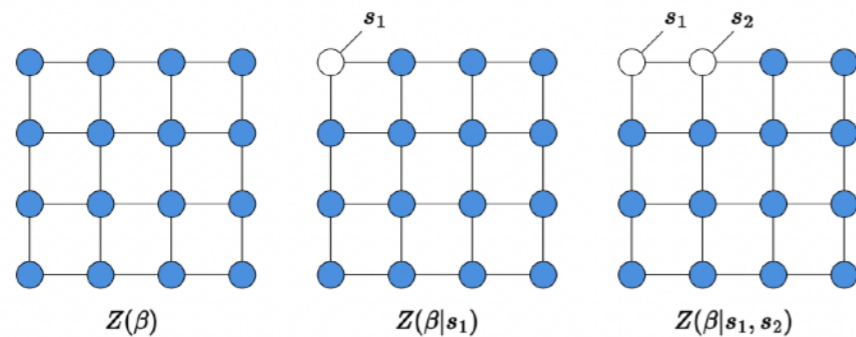
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TRG approaches

TN assisted Metropolis-Hastings  
collective updates



Frías-Pérez, Marien, Pérez-García, MCB, Iblisdir,  
SciPost Phys. 14, 123 (2023)

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as ansatz

TNS are very useful in the quantum  
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classify tensors (symmetries)

great descriptive power: phases,  
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TNS as (variational) ansätze for physical problems

efficient algorithms for GS, low excited states, thermal, dynamics

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**entanglement: crucial ingredient to understand QMB systems**

as (numerical) ansatz

# Tensor Network States (TNS)

efficient numerical algorithms (small spatial dimensions) and good theoretical understanding

non-technical review: [Annu Rev. CMP 2023 14:1](#);  
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new tools allow us to access some of these regimes

a question we want to address...

how do quantum systems thermalize?



# Thermalization of quantum systems

quantum system

$$H \quad |\Psi\rangle$$

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quantum system

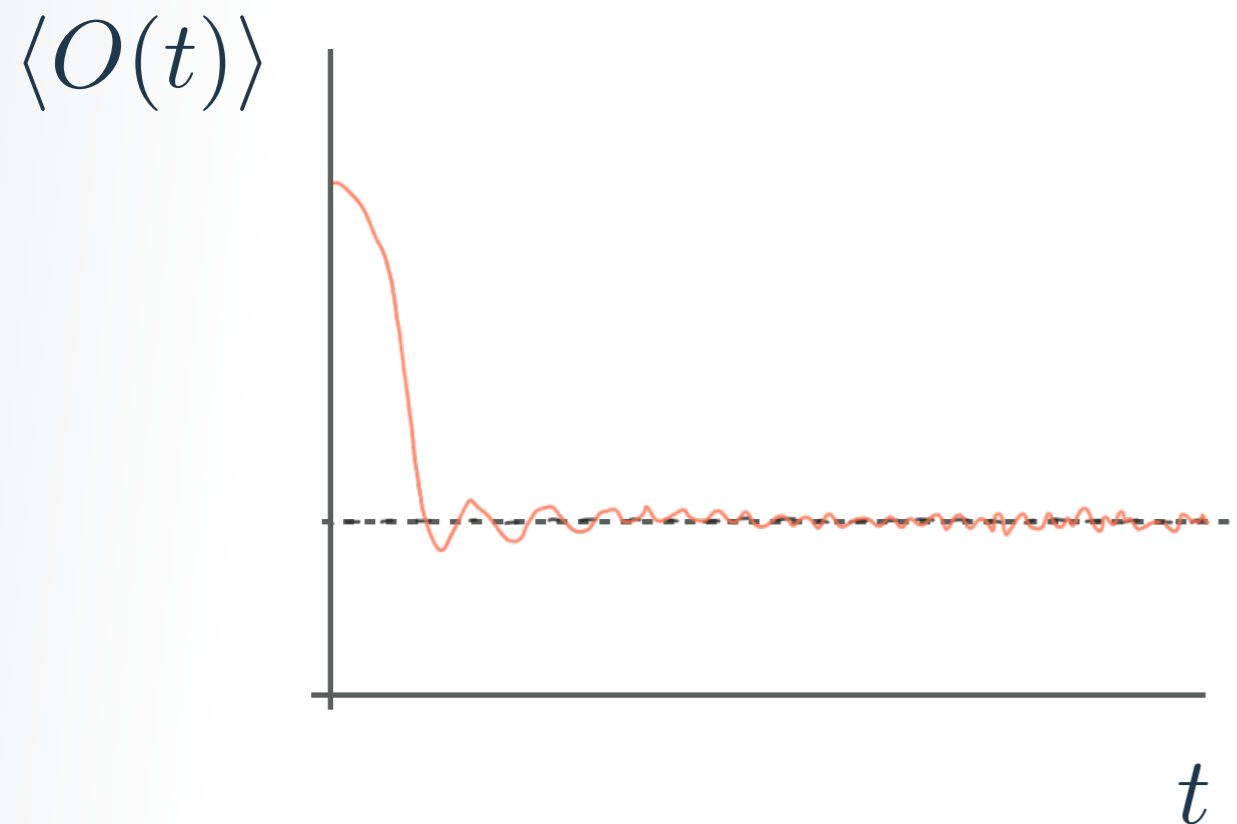
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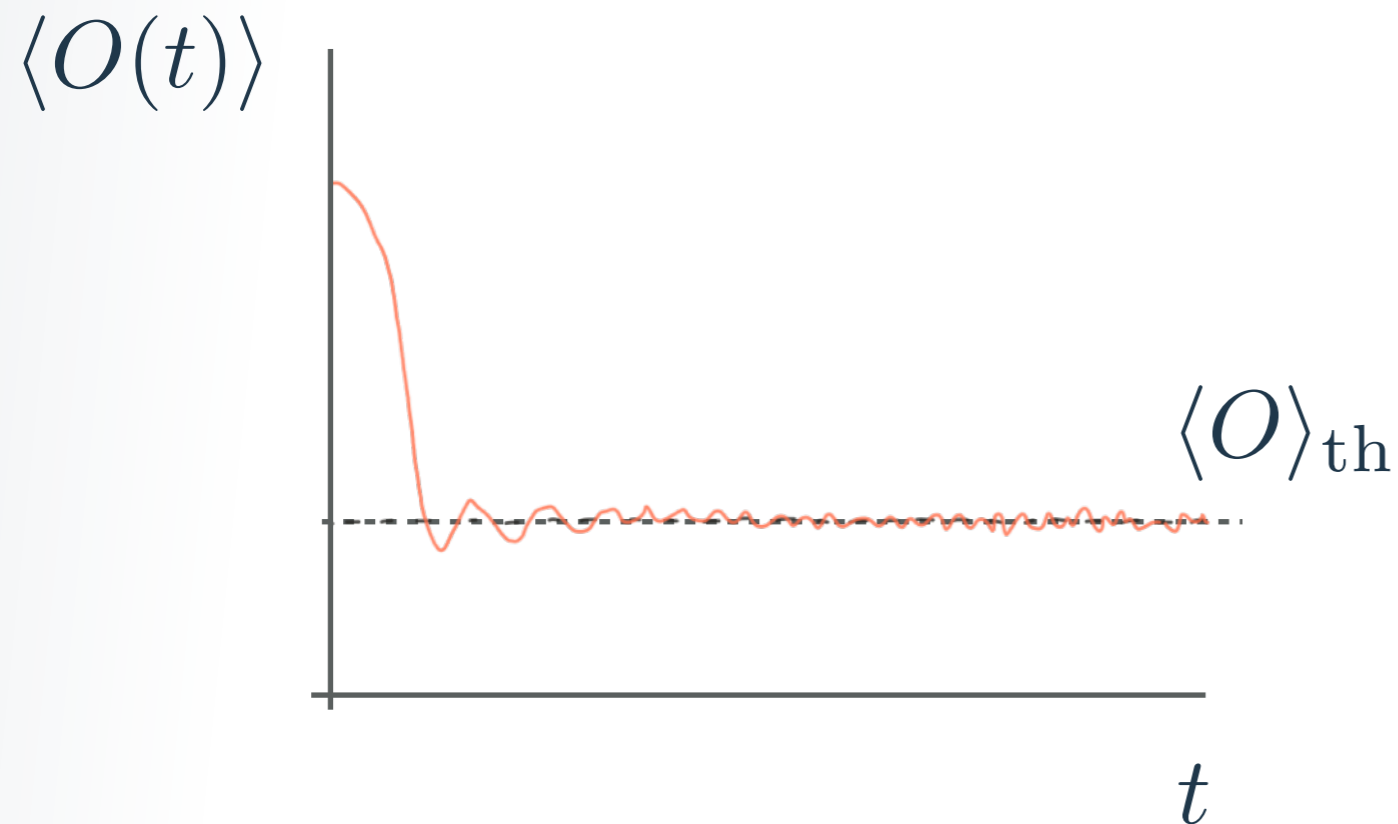


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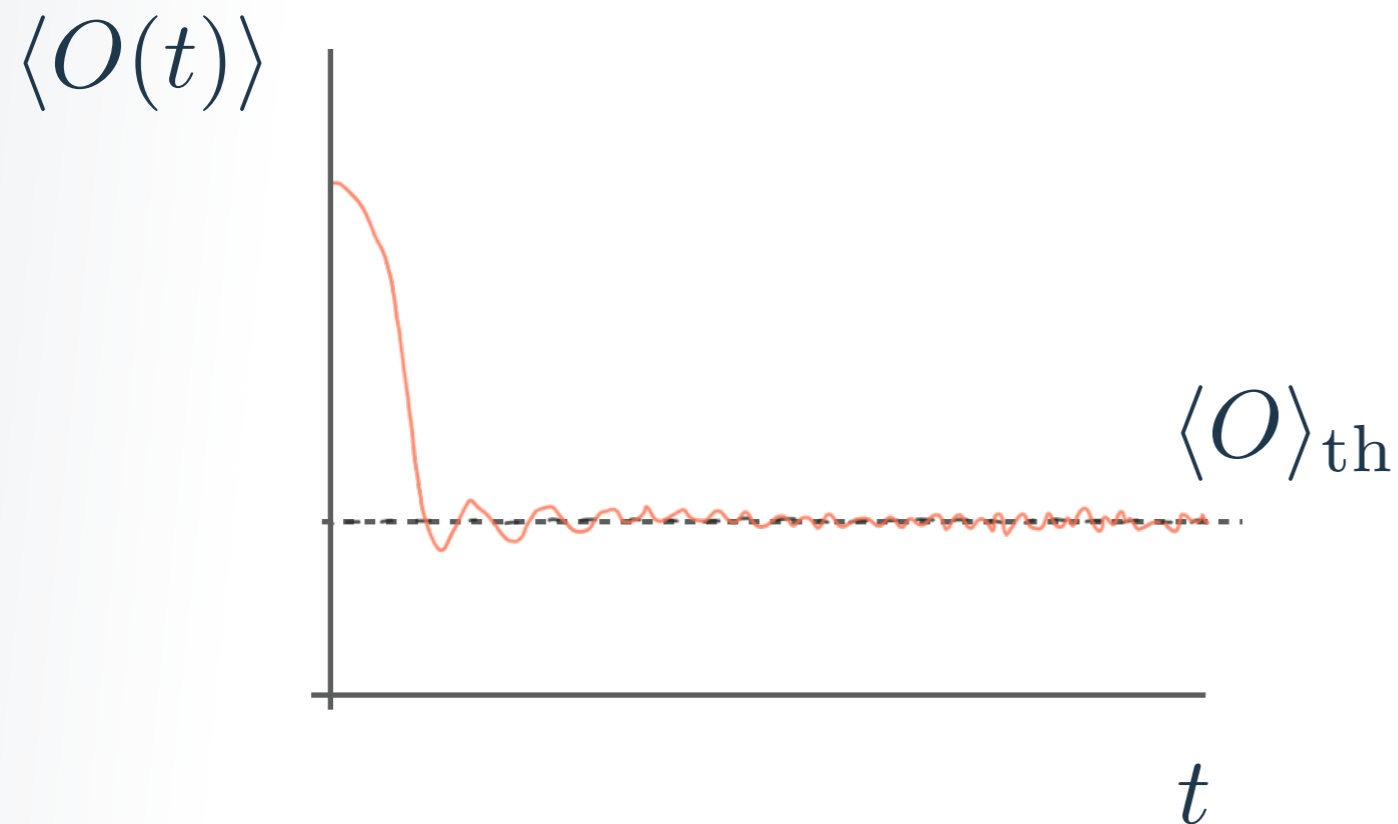
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reach (remain close to)  
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predicted by thermodynamic  
ensemble (microcanonical)

# Eigenstate Thermalization Hypothesis (ETH)

theoretical framework for quantum thermalization

review: D'Alessio et al, *Adv Phys* 65 (2016)

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ansatz for matrix elements of observables in energy  
eigenbasis

Srednicki, Deutsch 90s

$$O_{mn} = O(\bar{E})\delta_{mn} + e^{-\frac{S(\bar{E})}{2}} f_O(\bar{E}, \omega) R_{mn}$$

$$\bar{E} = \frac{E_m + E_n}{2}$$

$$\omega = E_m - E_n$$



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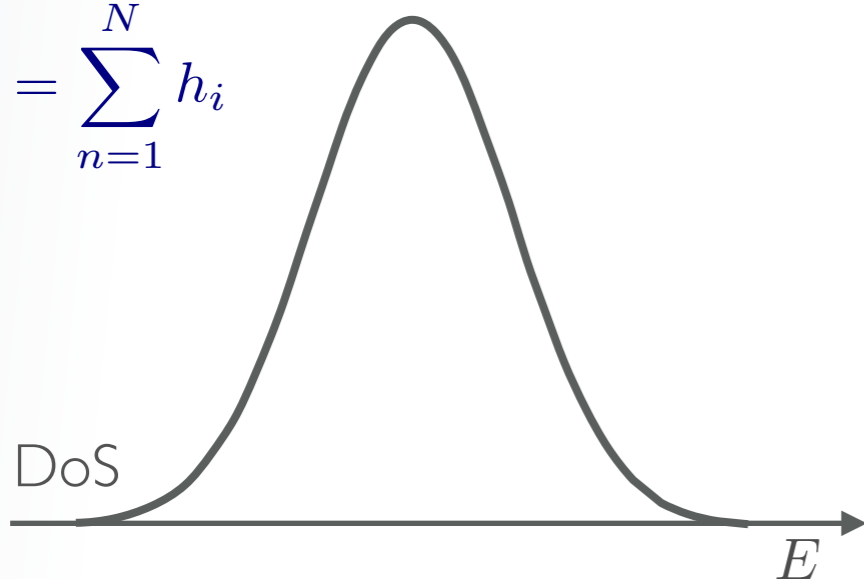
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implies (strong) thermalization for initial state  
with subextensive energy fluctuations

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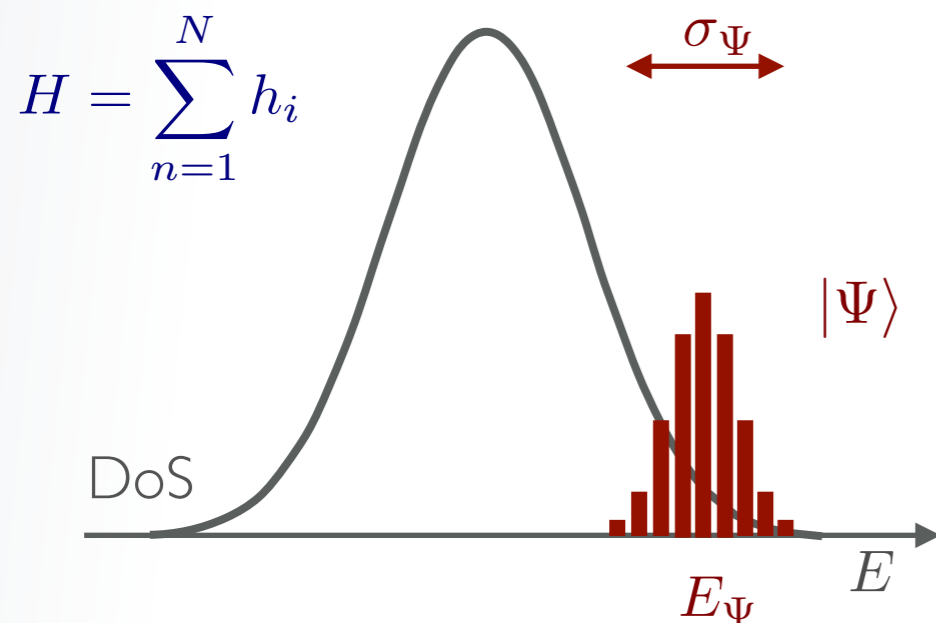
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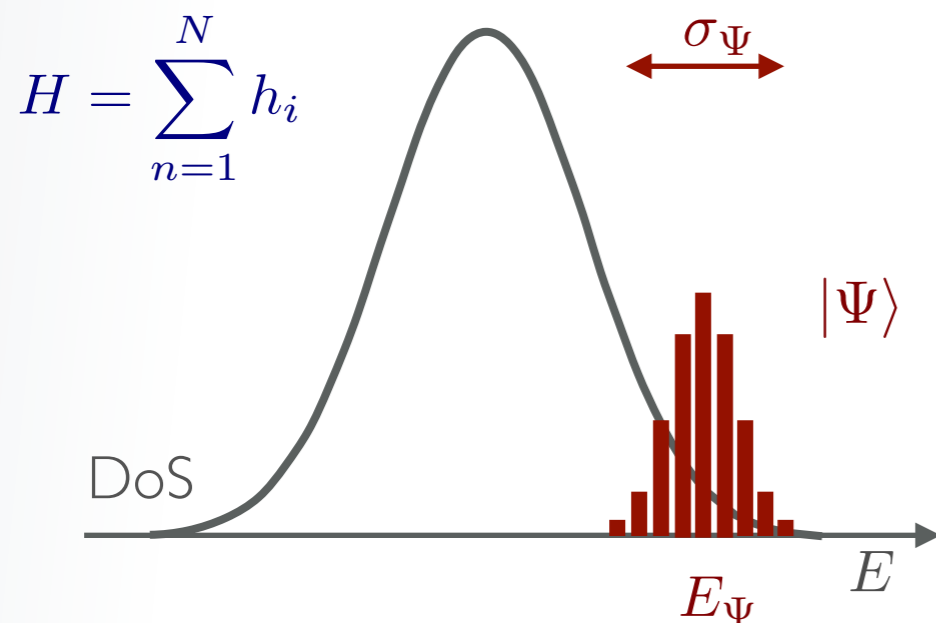
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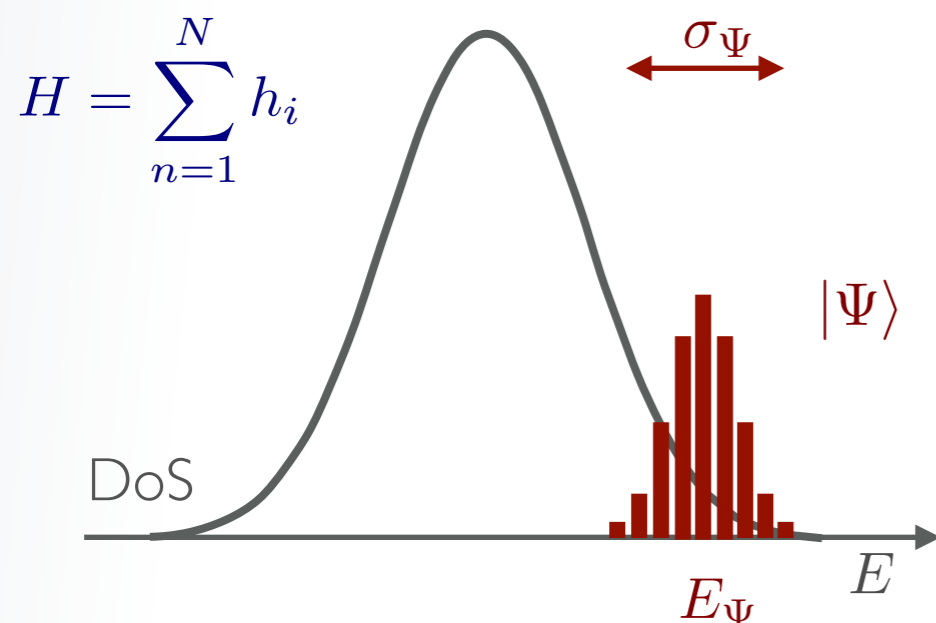
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fluctuations in time are  
exponentially small

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expected to hold for generic (non-integrable)  
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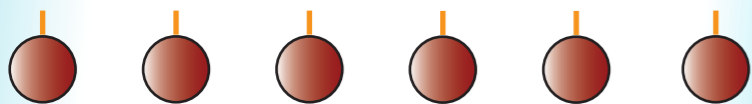
problem: exponentially large in  $N \Rightarrow$  Tensor Networks  
may be of help

simulation of non-equilibrium dynamics with MPS

# global quench in 1D

$t = 0$

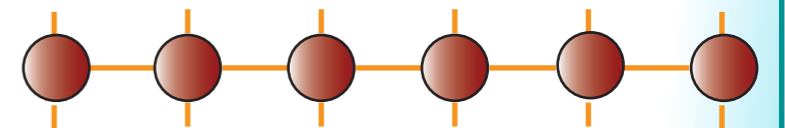
product state



easy to write as MPS

$t = \infty$

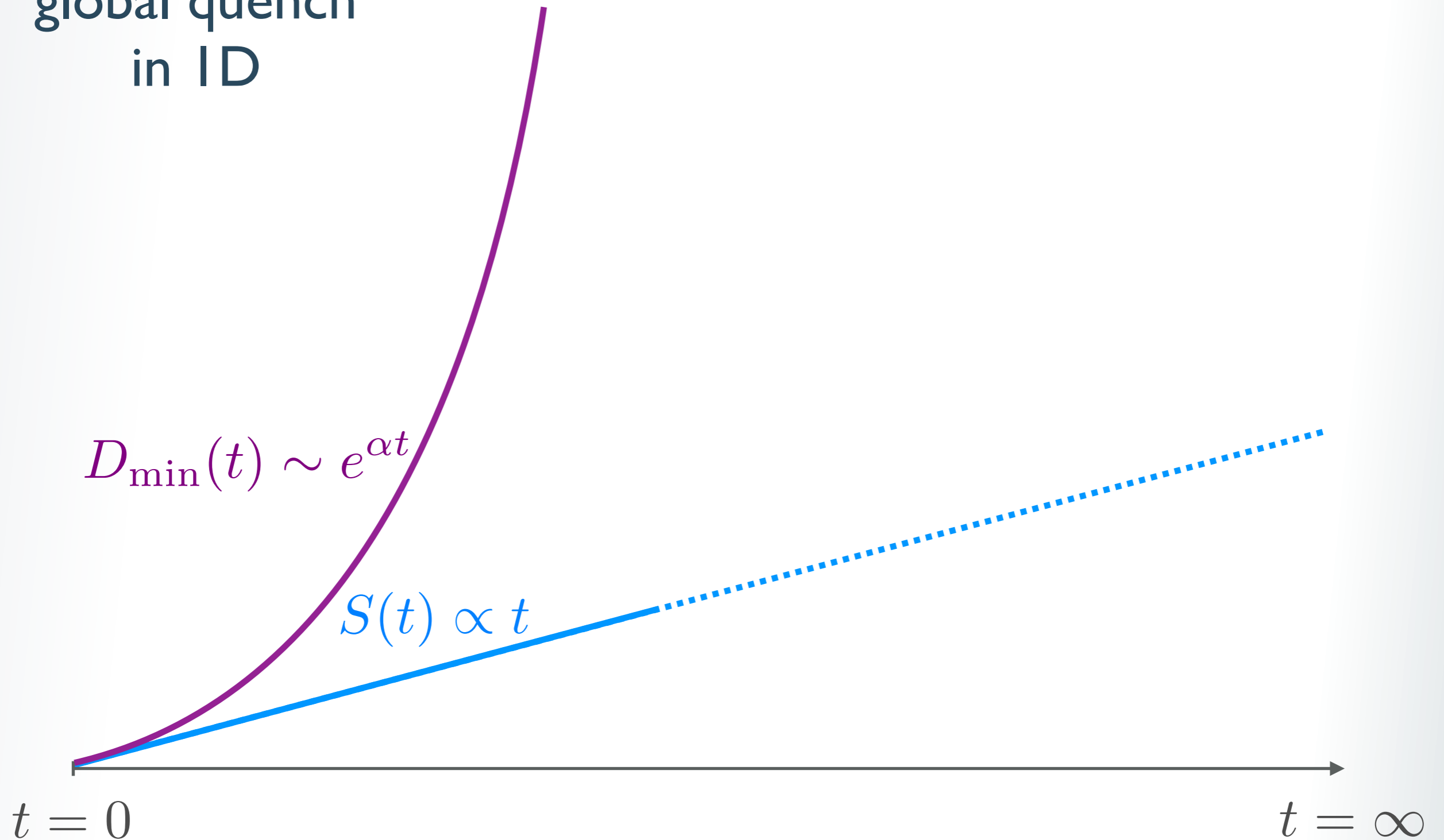
thermal states



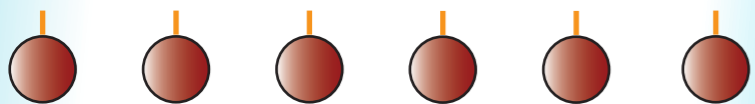
well approximated as MPO

local  
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# global quench in 1D



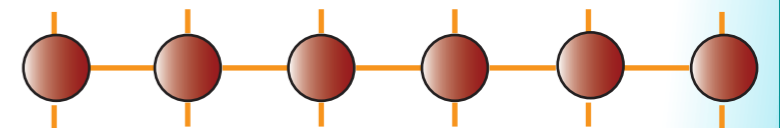
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# global quench in 1D

entanglement  
barrier

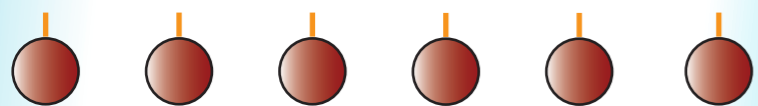
$$D_{\min}(t) \sim e^{\alpha t}$$

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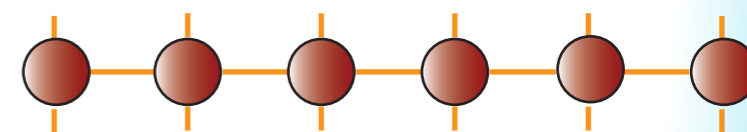
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global quench  
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TNS challenge:  
getting around this  
limitation

$$D_{\min}(t) \sim e^{\alpha t}$$

$$S(t) \propto t$$

ongoing effort

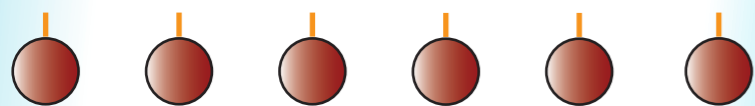
Dubail JPhysA 2017  
Leviatan et al. 2017  
White et al PRB 2018  
Surace et al. 2018  
Kvornig et al 2021  
Rakovzsky et al 2022

...

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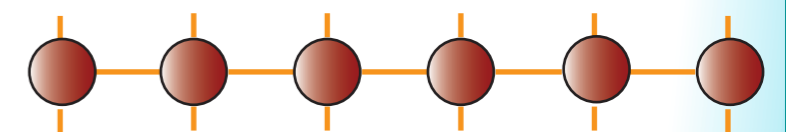
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local  
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thermal states



well approximated as MPO

our approach: an alternative  
quantum/TNS tool



# spectral (finite energy density) properties of the QMB Hamiltonian

MCB, Huse, Cirac, PRB 101, 144305 (2020)

Yang, Iblisdir, Cirac, MCB, PRL 124, 100602 (2020)

Papaefstathiou, Robaina, Cirac, MCB, PRD 104, 014514 (2021)

Çakan, Cirac, MCB, PRB 103, 115113 (2021)

**Lu, MCB, Cirac, PRX Quantum 2, 02032 (2021)**

**Yang, Cirac, MCB, PRB 106, 024307 (2022)**

**Luo, Trivedi, MCB, Cirac, PRB 109, 134304 (2024)**

# spectral (finite energy) properties

$$H = \sum_{n=1}^N h_n$$

$$\|h_n\| \leq 1$$

$$\text{tr } h_n = 0$$



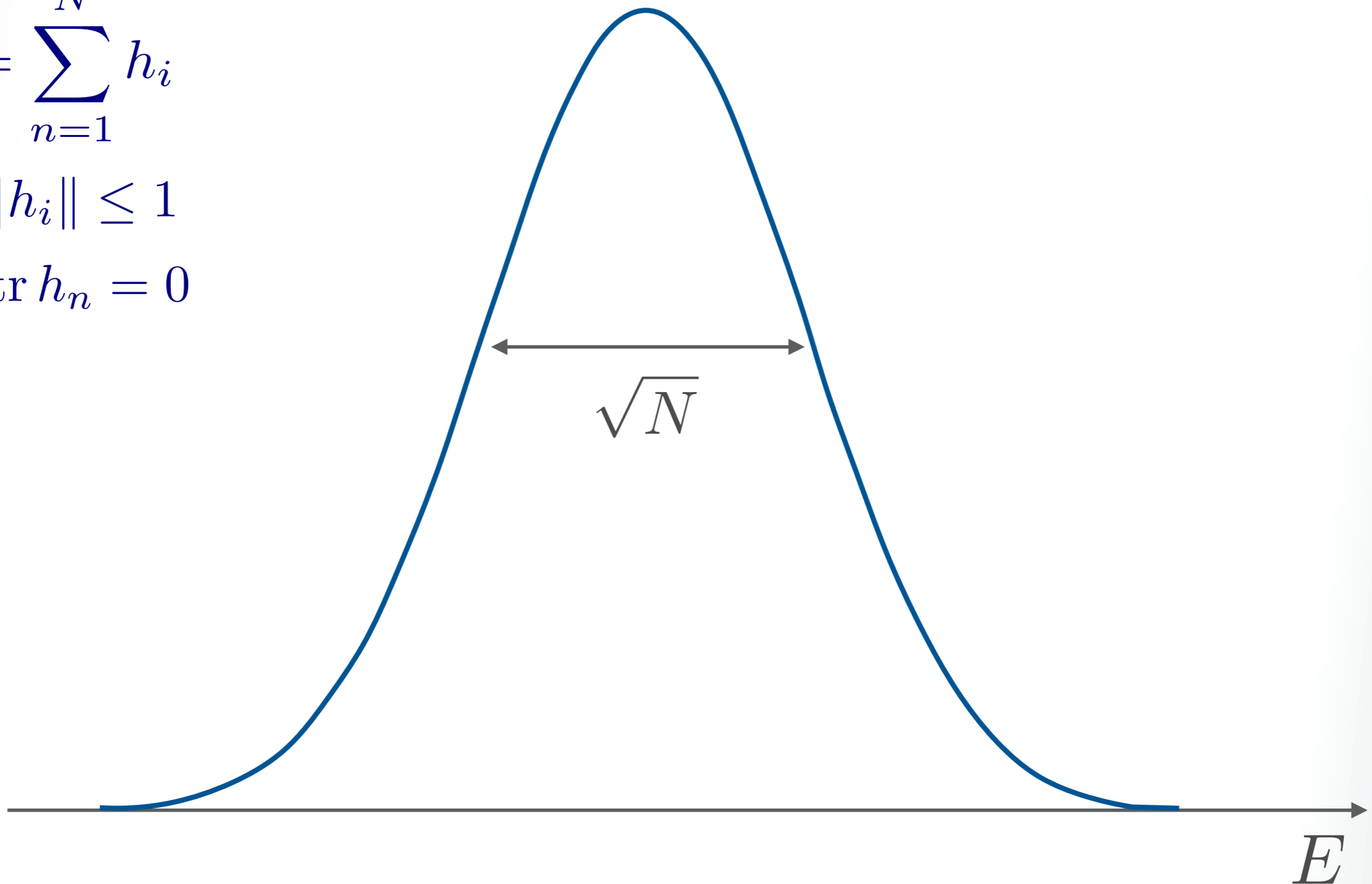
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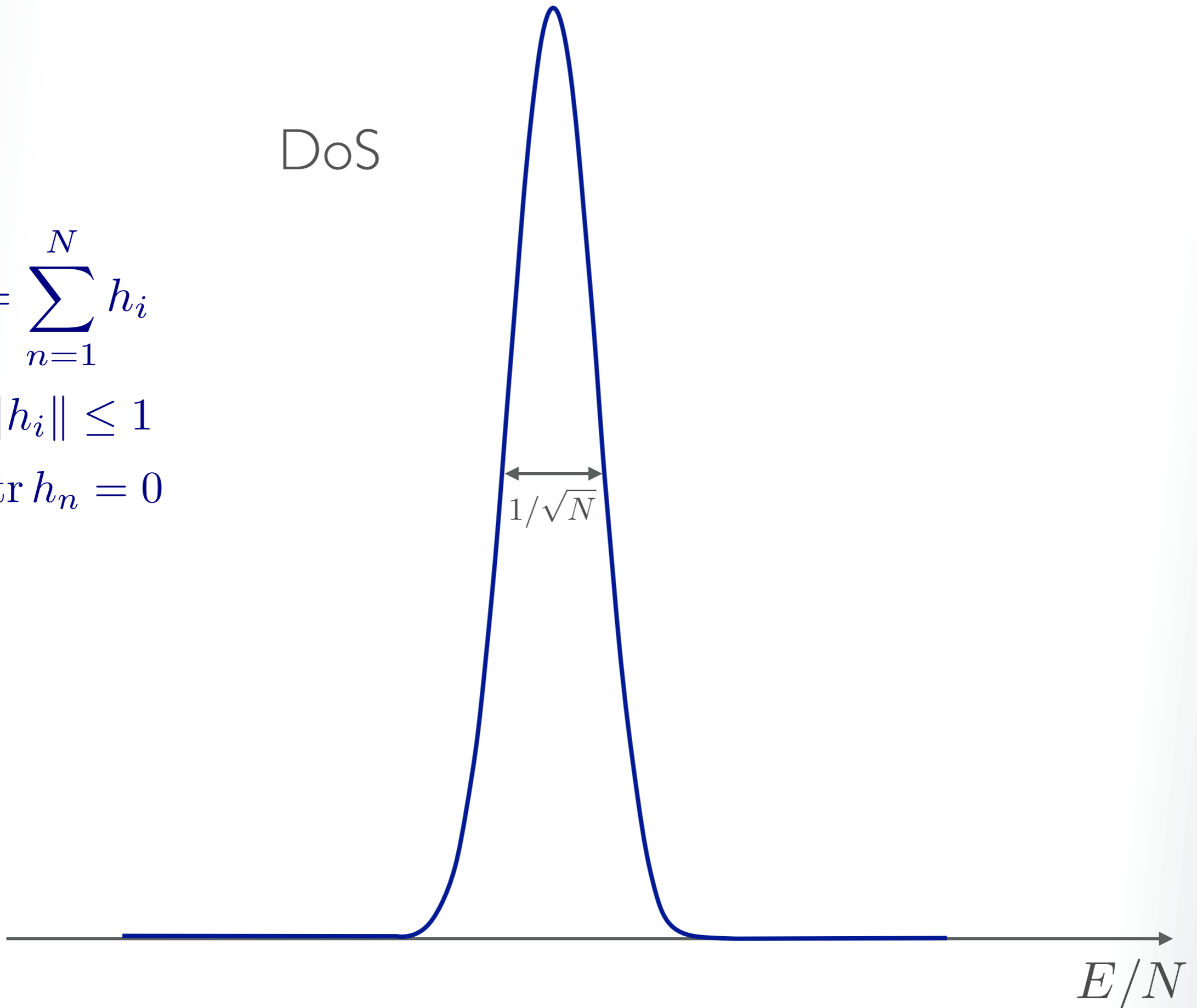
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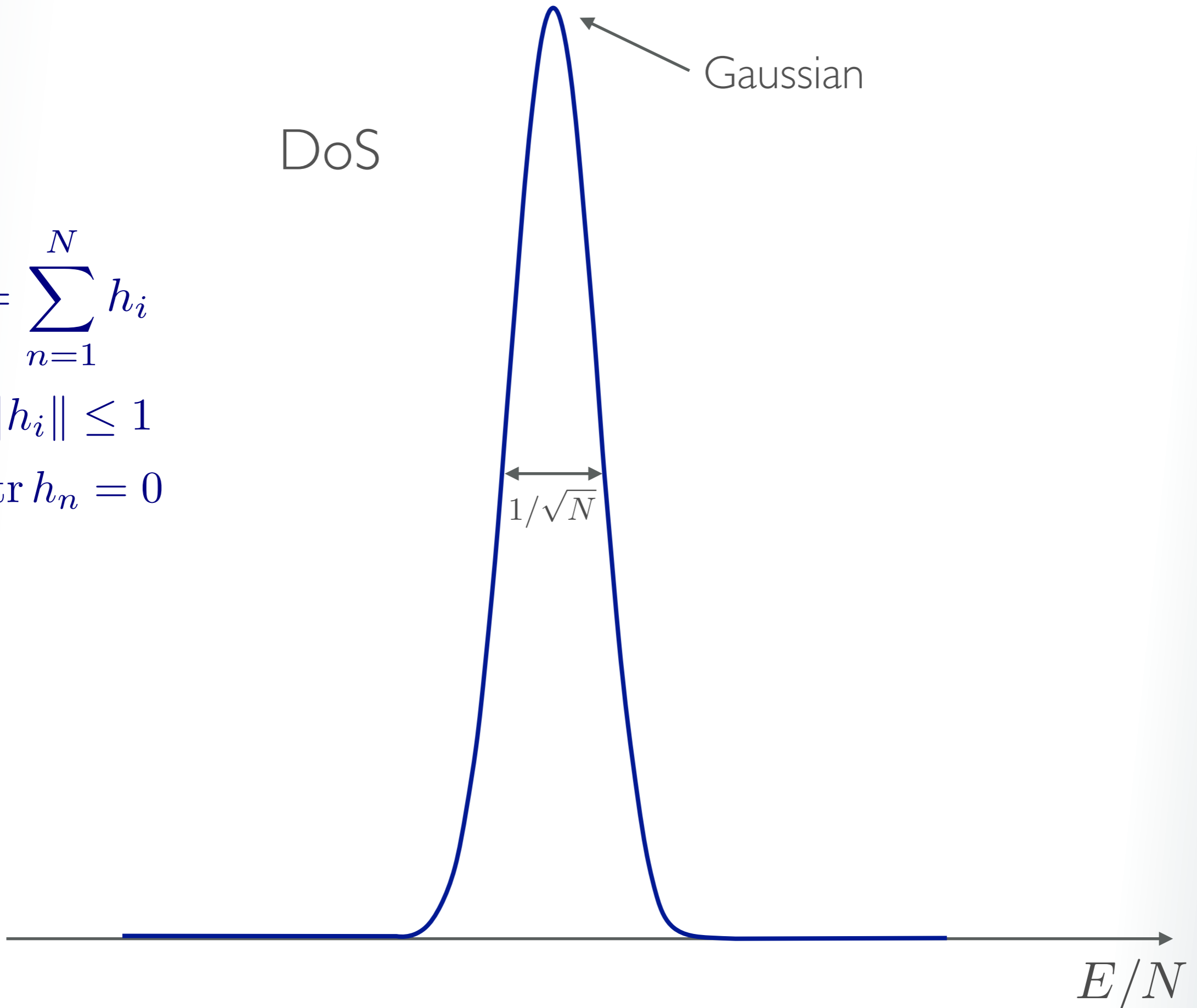
Gaussian

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$$1/\sqrt{N}$$



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$\hat{\delta}(H - E)$

$1/\sqrt{N}$

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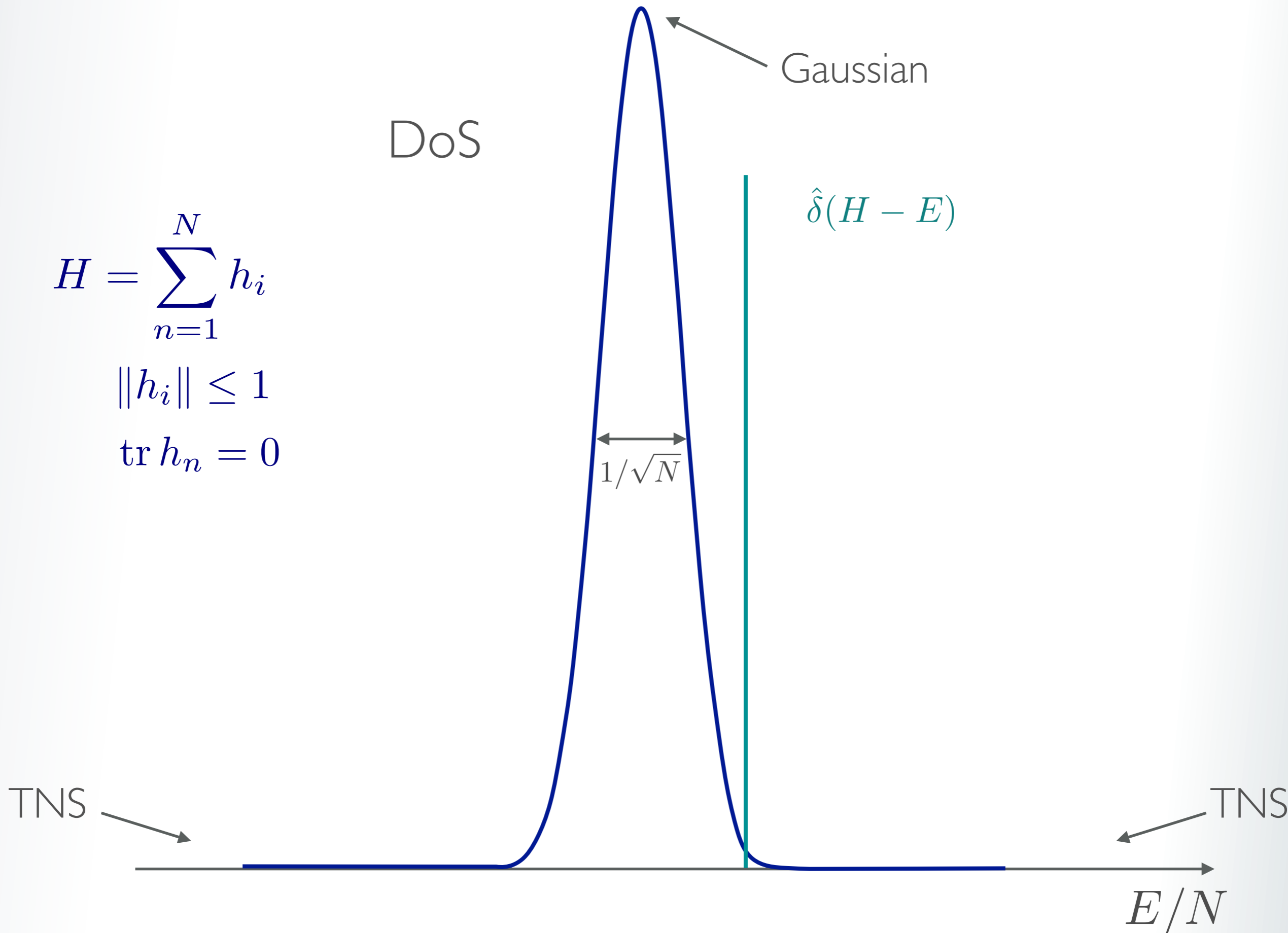
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$$\hat{\delta}(H - E)$$

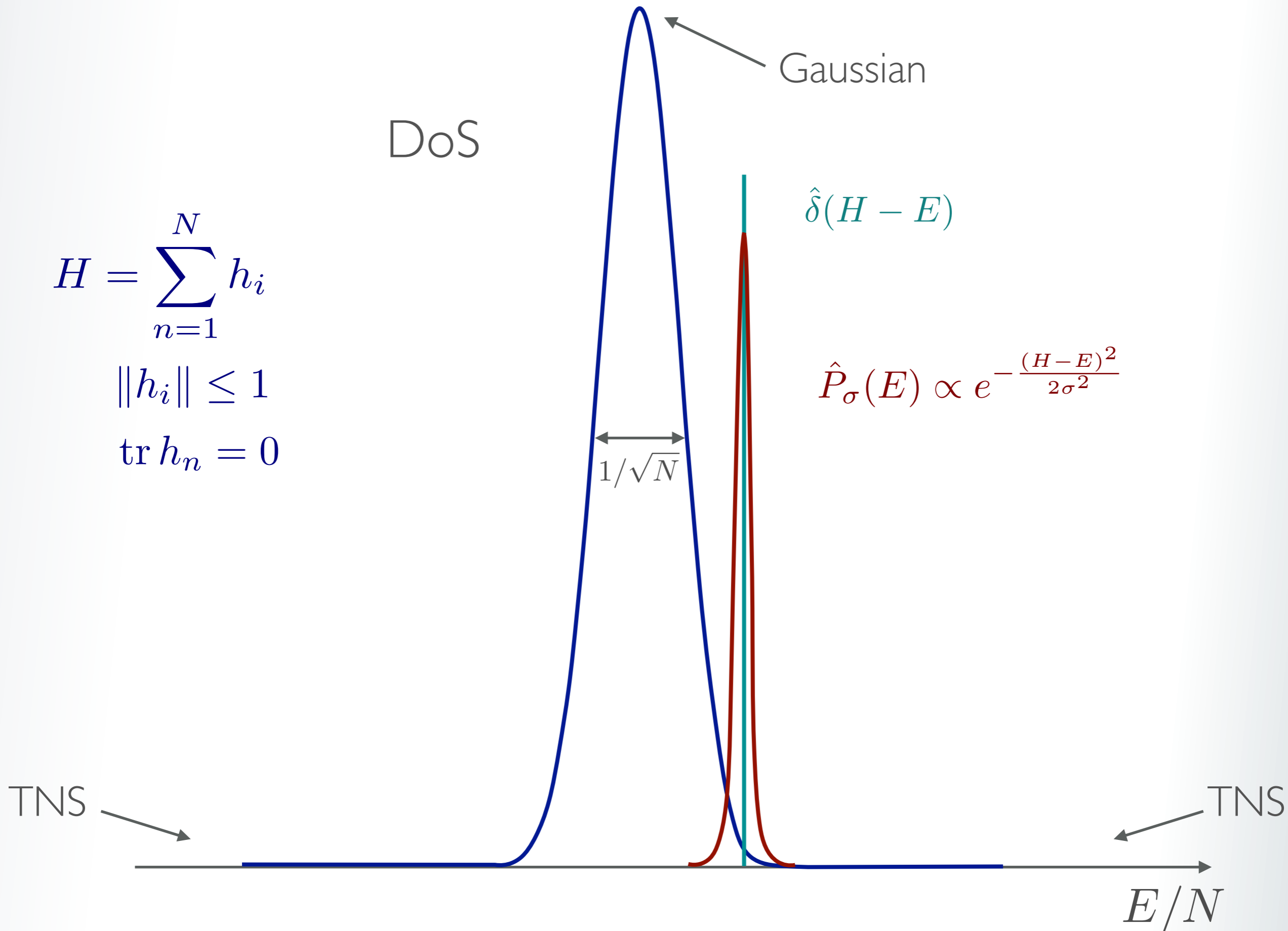
$$\hat{P}_\sigma(E) \propto e^{-\frac{(H-E)^2}{2\sigma^2}}$$

$$1/\sqrt{N}$$

TNS

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$E/N$





## generalized density of states

$$g(E; O) = \sum_n \delta(E - E_n) \langle E_n | O | E_n \rangle$$

microcanonical average:  
expectation value in eigenstates  
with given energy

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unnormalised DOS

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unnormalised DOS

$$O = |\Psi\rangle\langle\Psi| \quad \longrightarrow \quad g(E; \Psi) = \sum_n \delta(E - E_n) |\langle E_n | \Psi \rangle|^2$$

local DOS

# implementing the filter

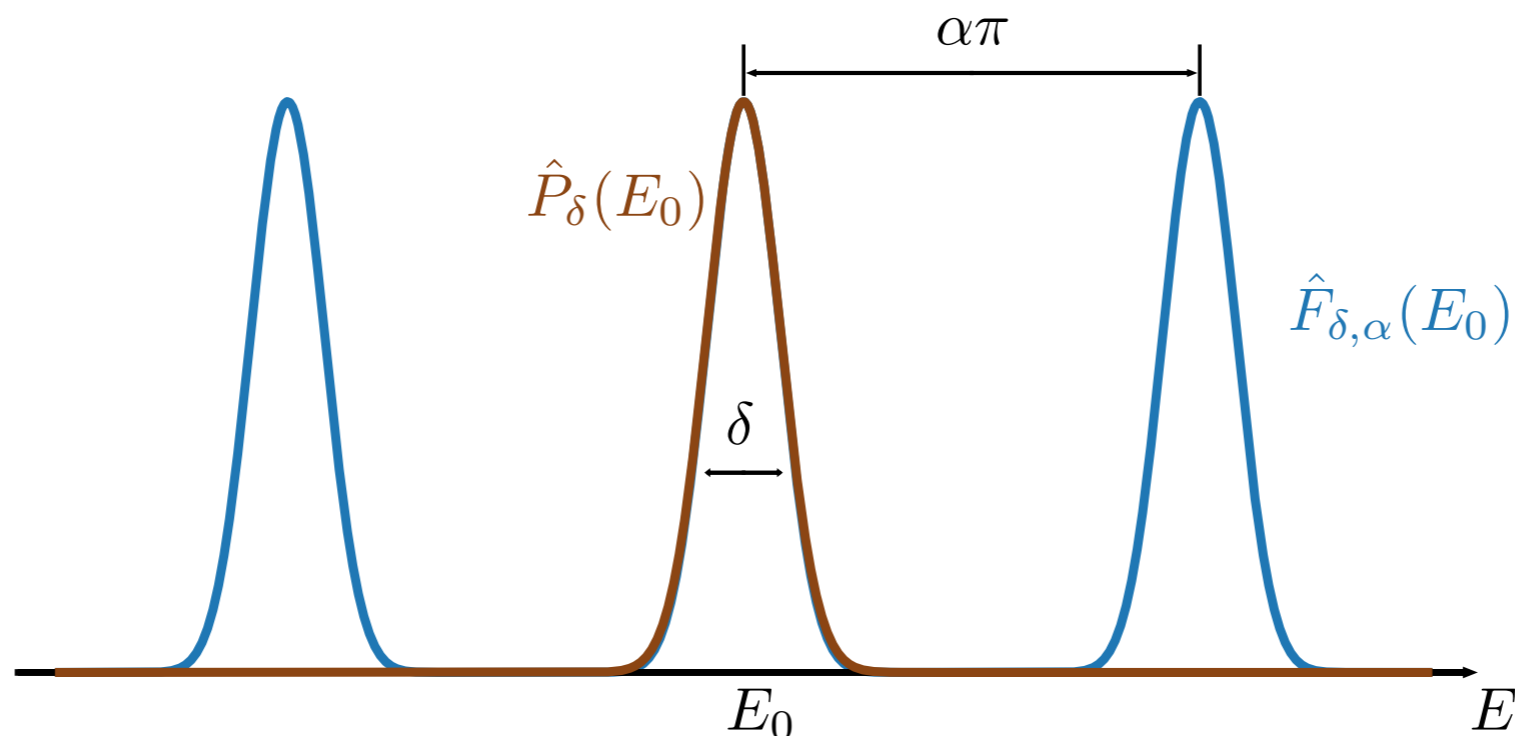
Gaussian operator  $\Rightarrow$  not local

# implementing the filter

Gaussian operator  $\Rightarrow$  not local

$\Rightarrow$  cosine approximation

$$\cos^M(x) \approx e^{-Mx^2/2} \quad x < \pi/2$$



# implementing the filter

Gaussian operator  $\Rightarrow$  not local  
 $\Rightarrow$  cosine approximation

$$[\cos X]^M =$$

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$$[\cos X]^M = \frac{1}{2^M} \sum_{m=-M/2}^{M/2} \binom{M}{M/2 - m} e^{i2mX}$$

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peak at  $m=0$

cut the sum  $\propto \sqrt{M}$



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Gaussian operator  $\Rightarrow$  not local

$\Rightarrow$  cosine approximation

$$[\cos X]^M = \frac{1}{2^M} \sum_{m=-M/2}^{M/2} \binom{M}{M/2 - m} e^{i2mX}$$

peak at  $m=0$

cut the sum  $\propto \sqrt{M}$

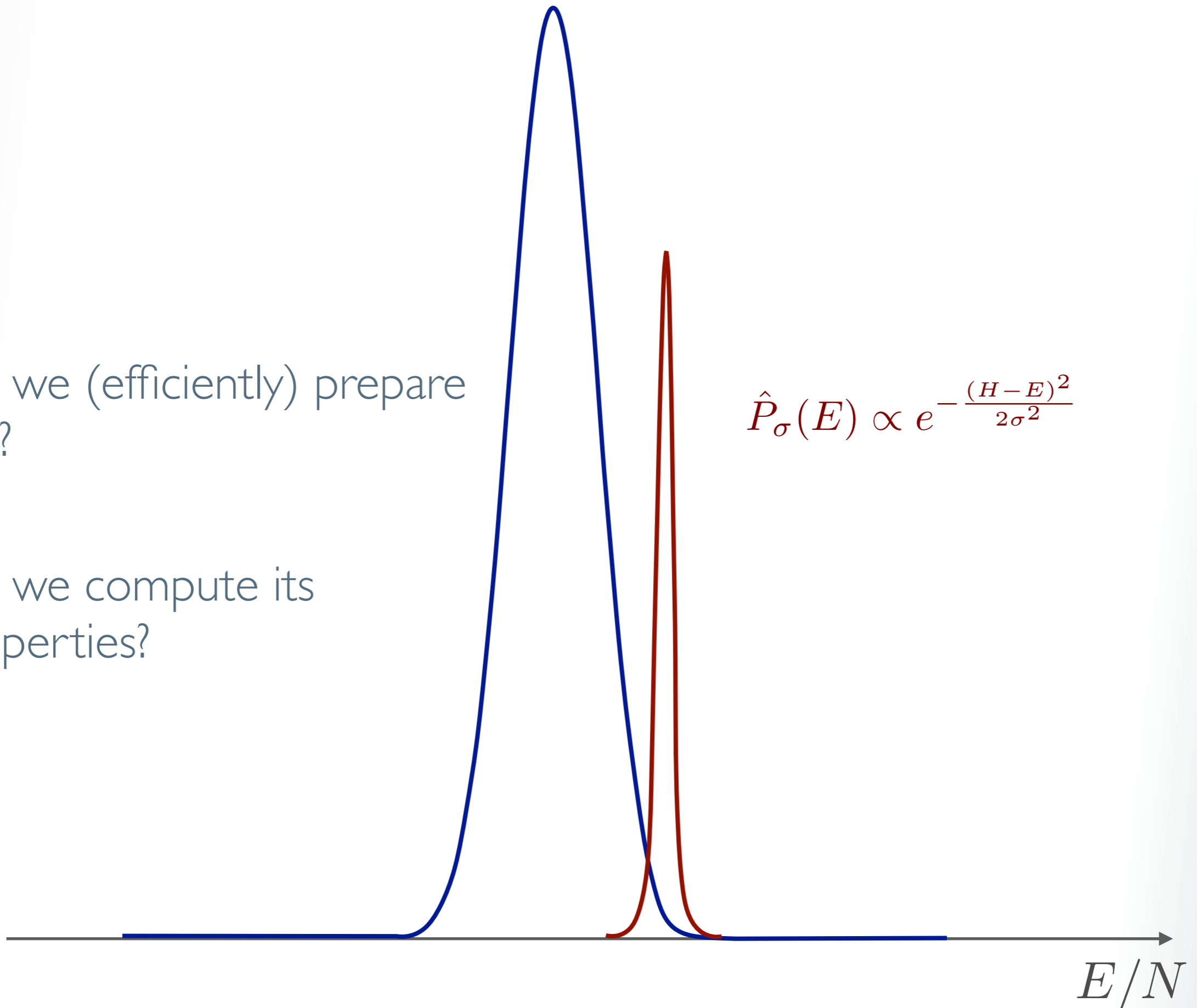
$$\exp\left(-\frac{H^2}{2\delta^2}\right) \approx \sum_{m=-x\alpha/\delta}^{x\alpha/\delta} c_m e^{i2mH/\alpha}$$

This talk

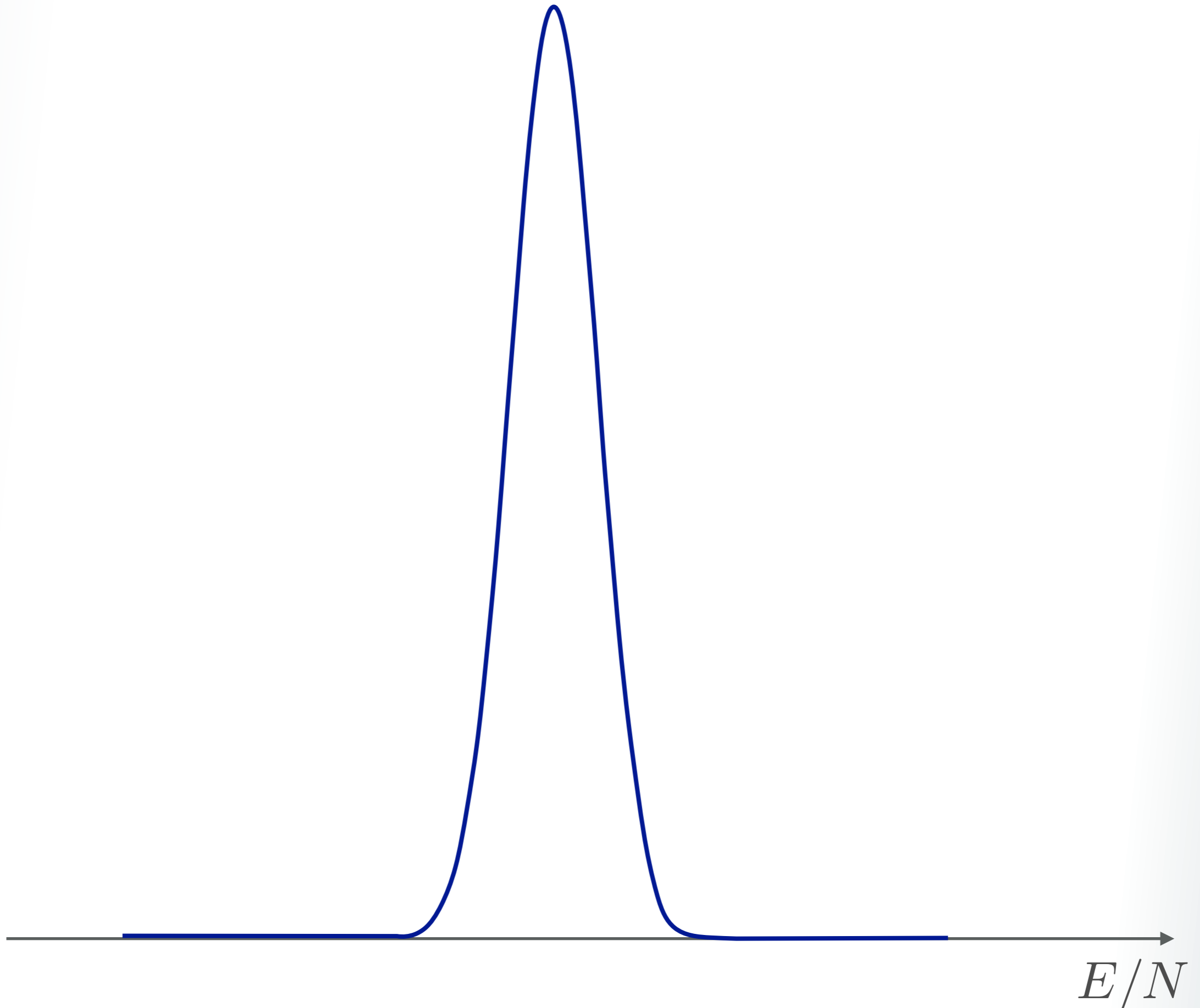
can we (efficiently) prepare this?

can we compute its properties?

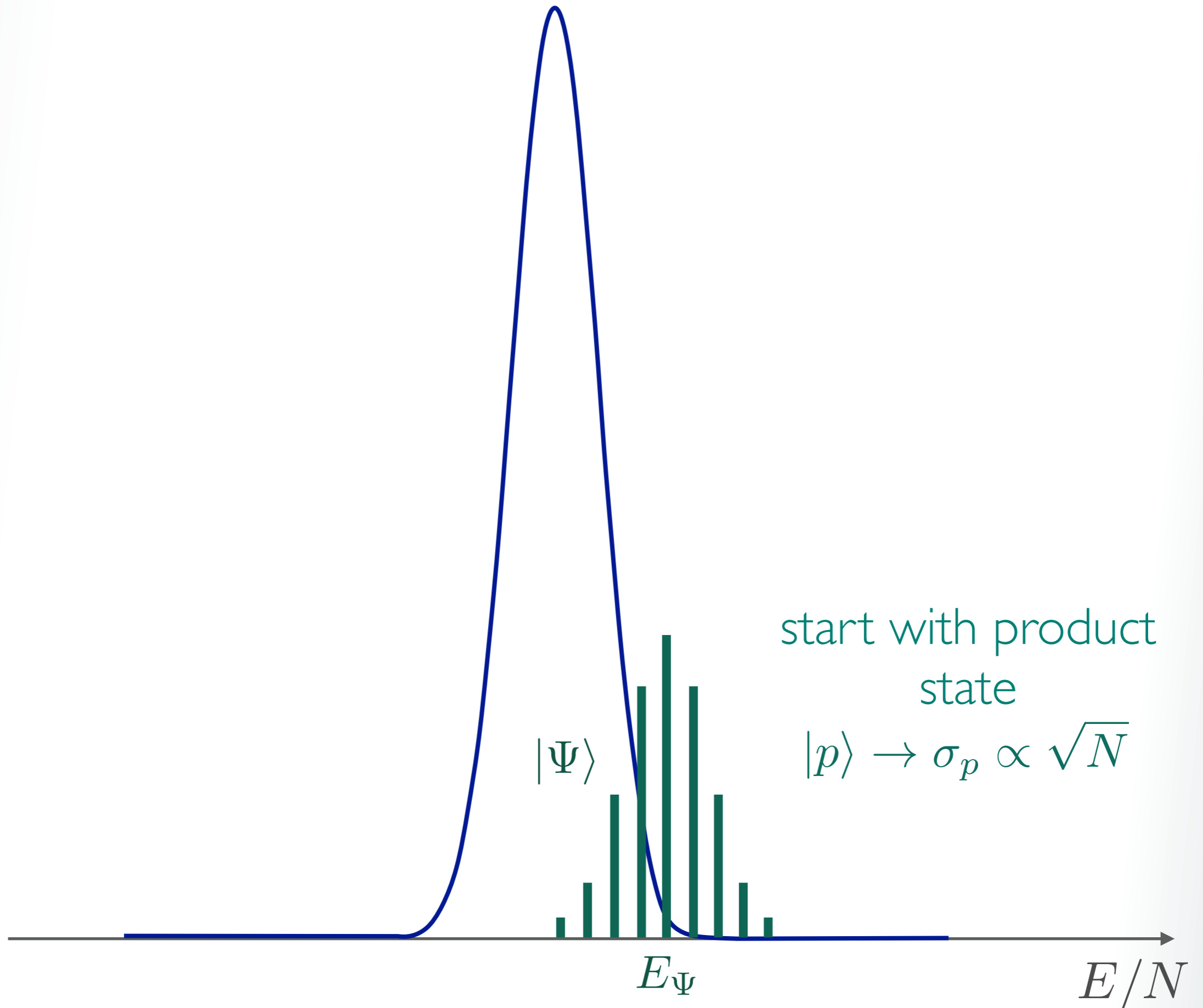
$$\hat{P}_\sigma(E) \propto e^{-\frac{(H-E)^2}{2\sigma^2}}$$



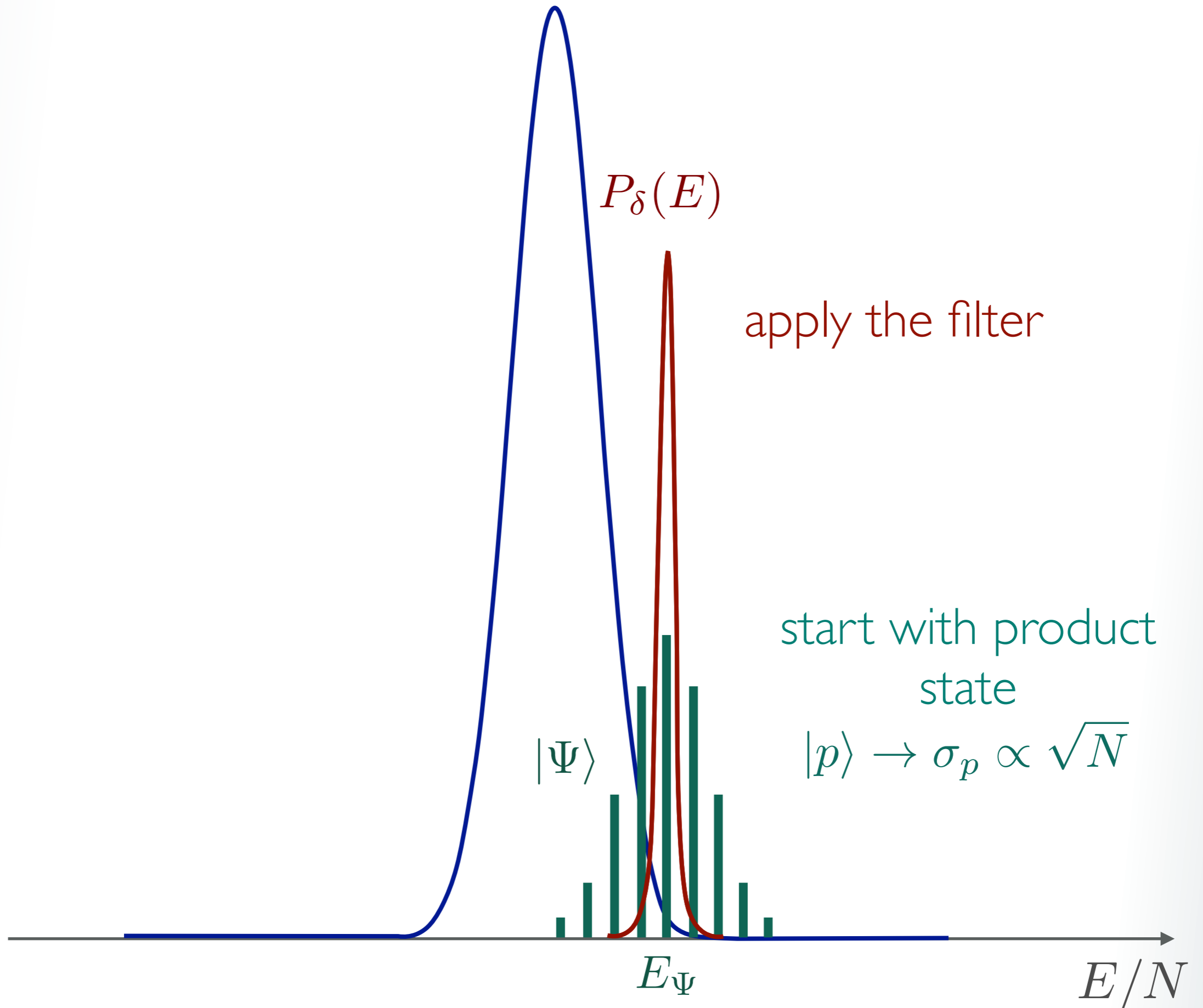
Can we prepare a filtered state?



Can we prepare a filtered state?



Can we prepare a filtered state?



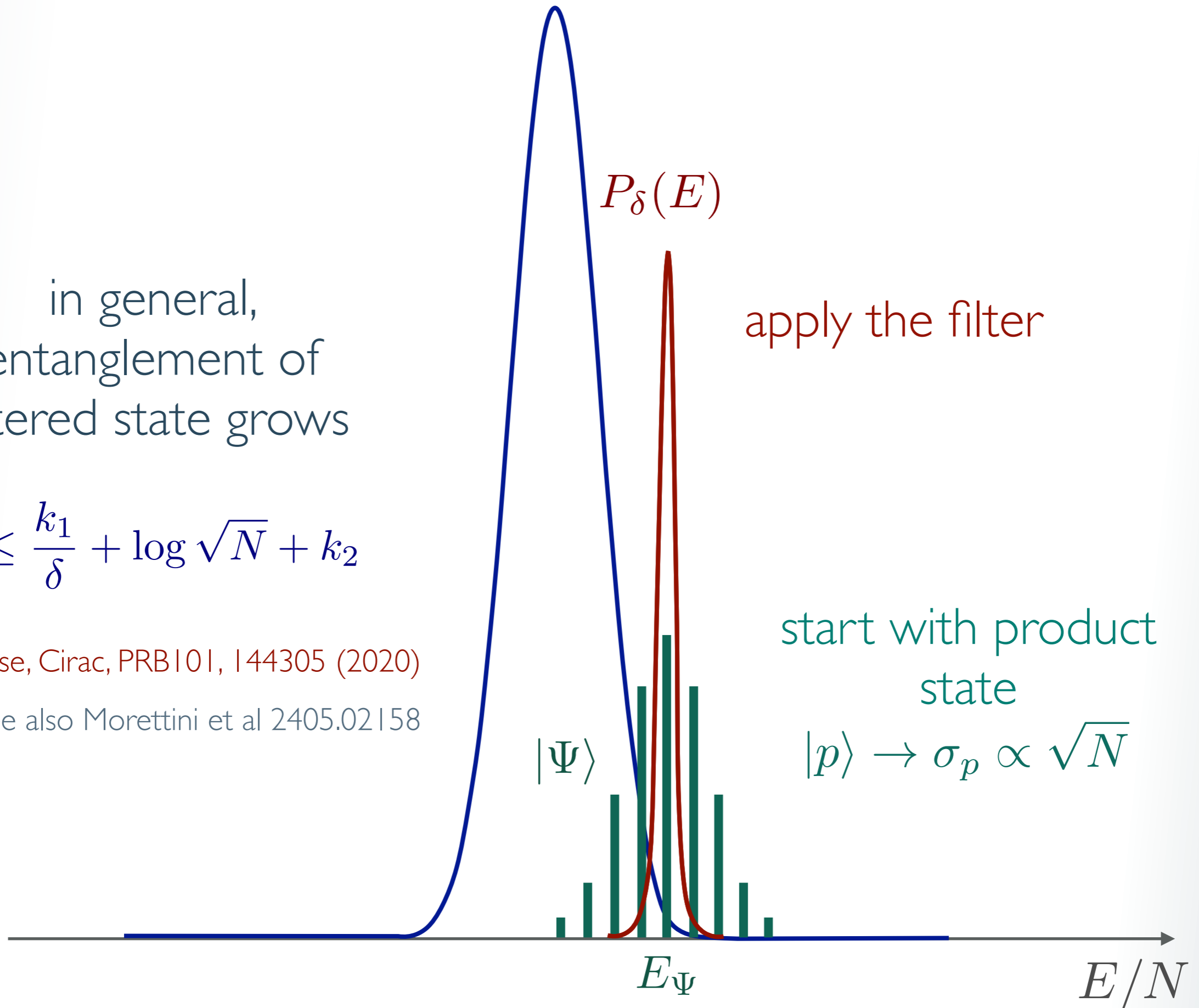
# Can we prepare a filtered state?

in general,  
entanglement of  
filtered state grows

$$S \leq \frac{k_1}{\delta} + \log \sqrt{N} + k_2$$

MCB, Huse, Cirac, PRB 101, 144305 (2020)

see also Morettini et al 2405.02158



# reducing the energy variance costs entanglement

IDEA:

$$\exp \left[ -\frac{(H - E)^2}{2\delta^2} \right] |p\rangle \approx \sum_{m=-R}^R c_m e^{-i2mE/\alpha} e^{i2mH/\alpha} |p\rangle$$

# reducing the energy variance costs entanglement

IDEA:

I. reduce the number of terms (overlap)

Hartmann et al., Lett. Math. Phys. 2004

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2. each term bounded  $S \Rightarrow D$

van Acoleyen et al. PRL 2013

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van Acoleyen et al. PRL 2013

3. sum has bounded  $D \Rightarrow S$

$$D \leq C\sqrt{N} \left( D_0^{1/\delta} - 1 \right)$$

$$S \leq \frac{k_1}{\delta} + \log \sqrt{N} + k_2$$

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Hartmann et al., Lett. Math. Phys. 2004

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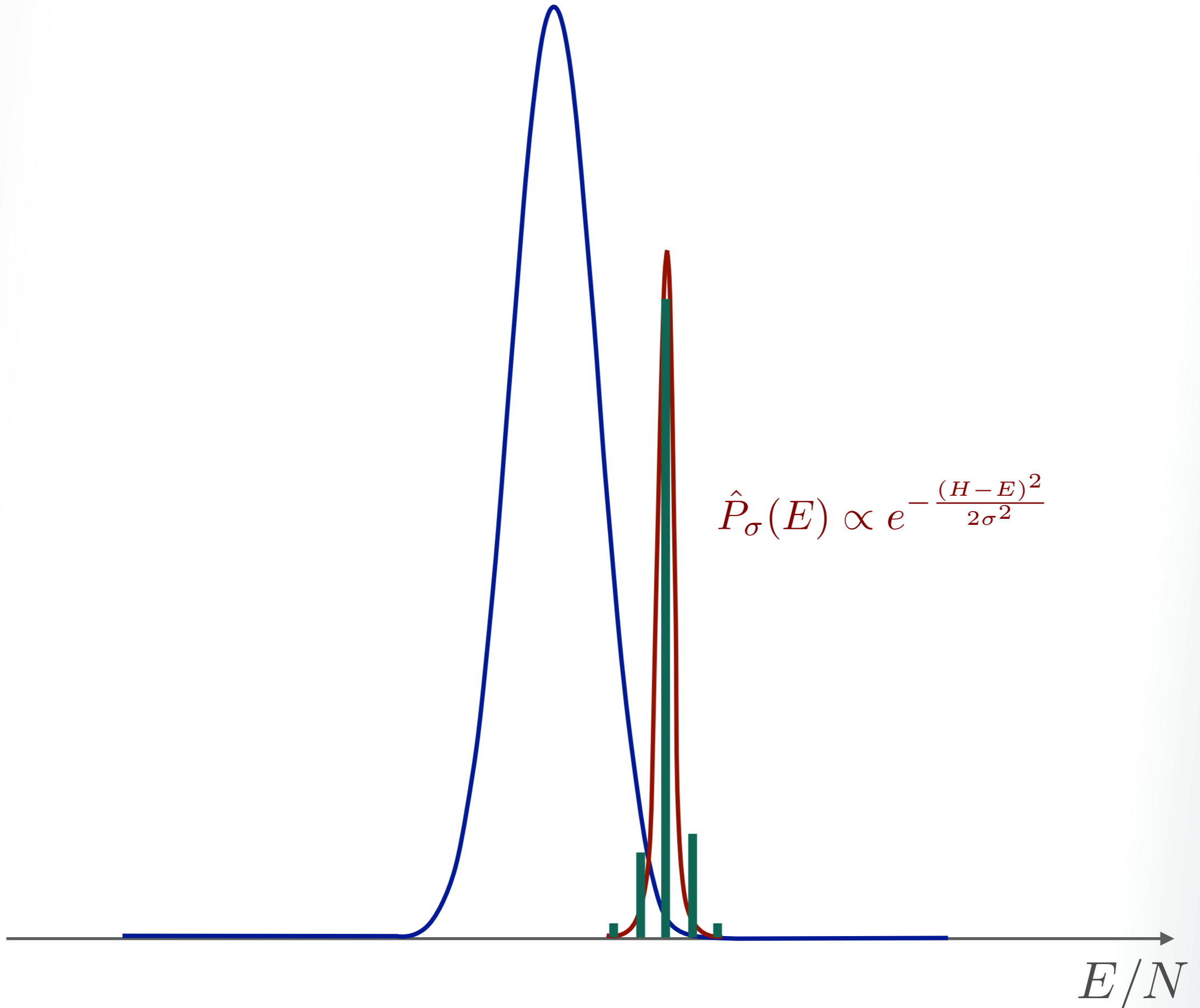
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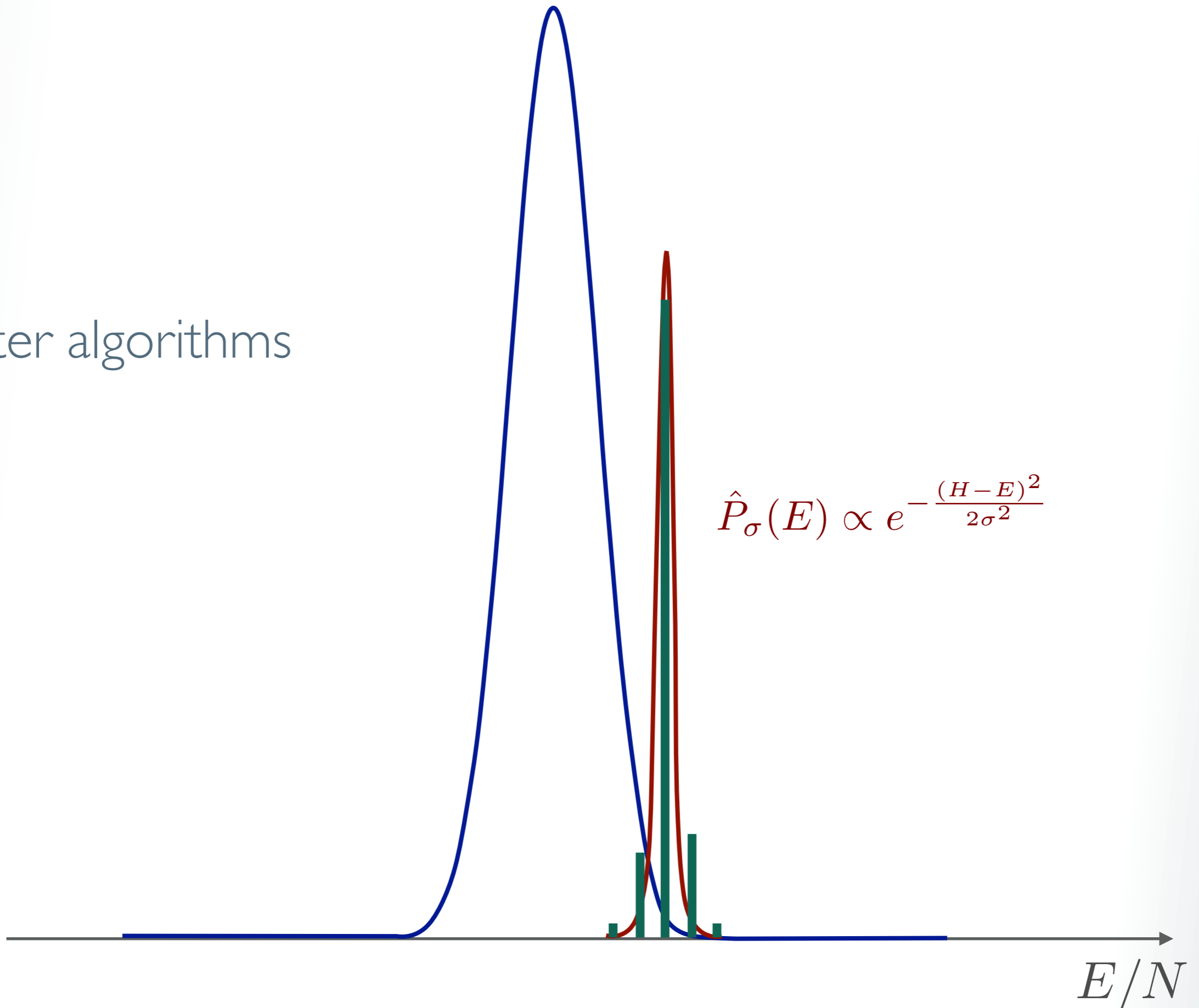
MPS could reach  
 $\delta^{-1} \sim O(\log N)$

Can we compute the properties of this state?



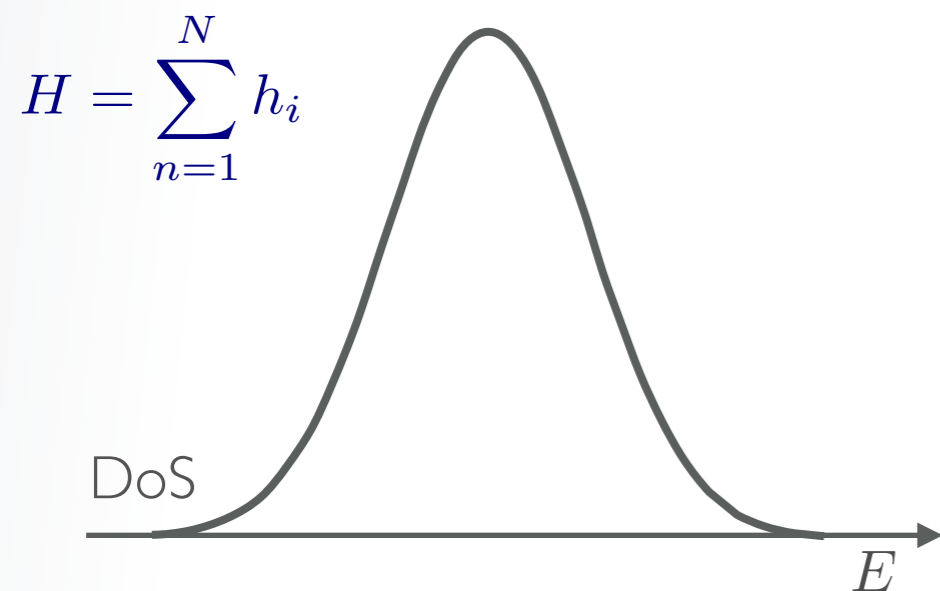
Can we compute the properties of this state?

filter algorithms



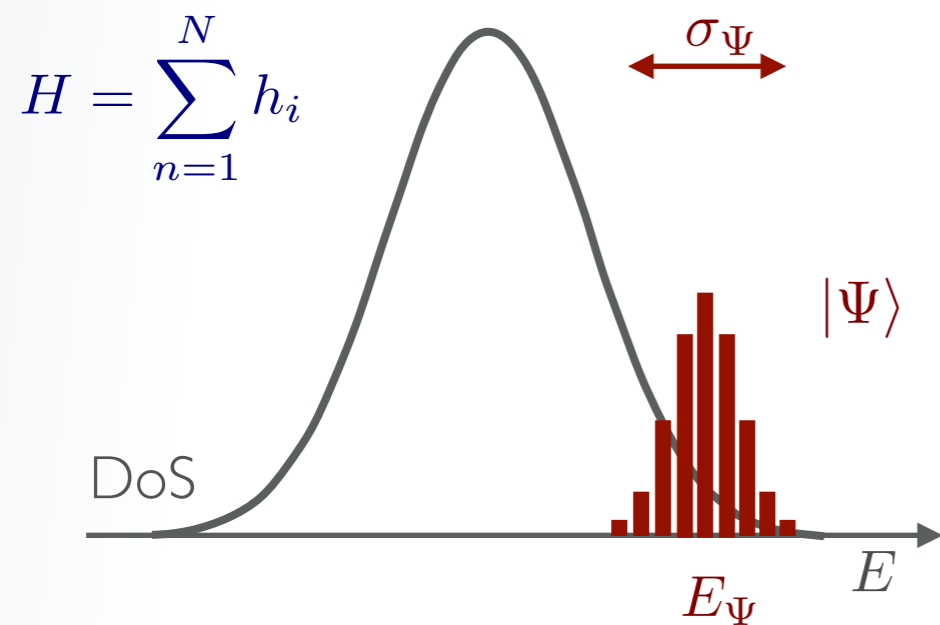
# energy filter

## 1 filtering a state



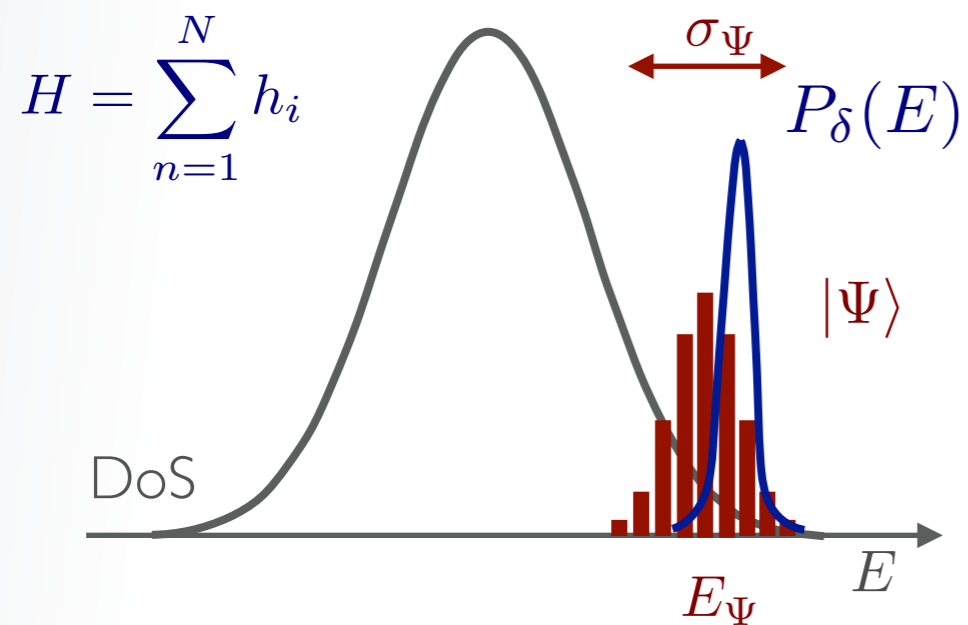
# energy filter

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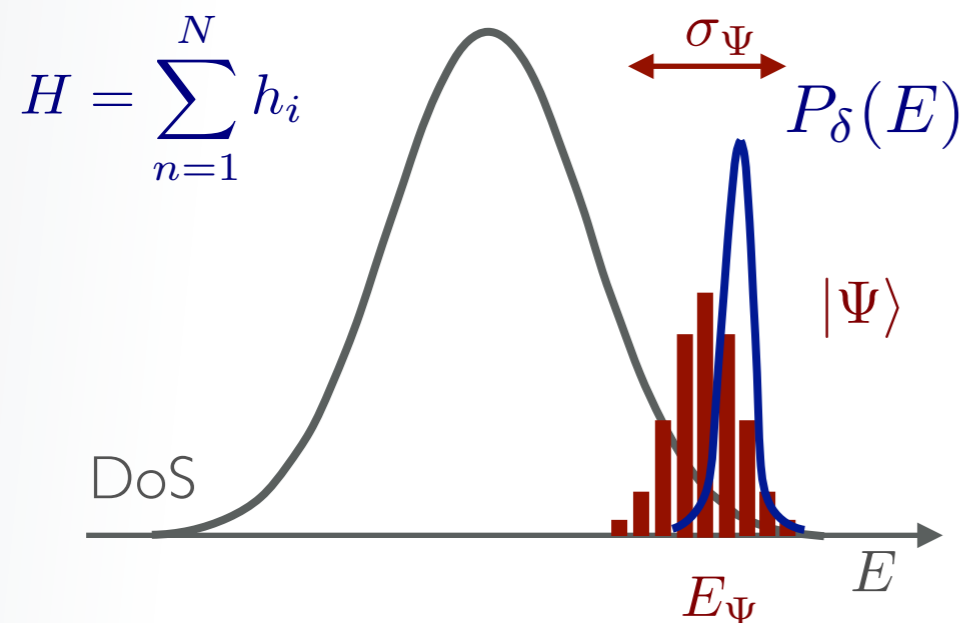
# energy filter

## 1 filtering a state

decrease energy variance  $\Rightarrow$  microcanonical

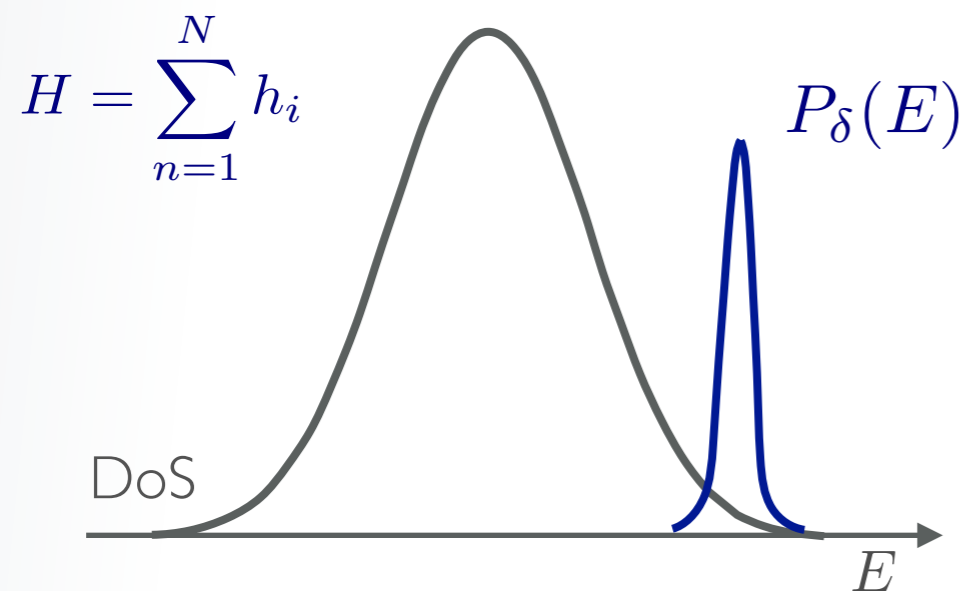
$$\langle P_\delta(E)\Psi | O | P_\delta(E)\Psi \rangle \Rightarrow O(E)$$

$$\langle \Psi | P_\delta(E) | \Psi \rangle \Rightarrow \text{LDOS}$$



# energy filter

## 2 filter as ensemble



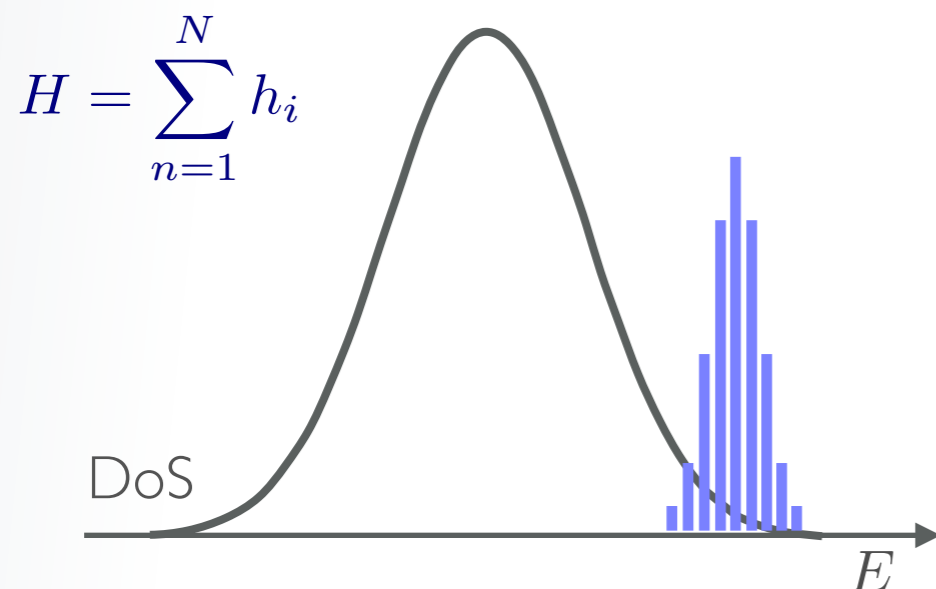
# energy filter

## 2 filter as ensemble

diagonal in energy eigenbasis  $\Rightarrow$  microcanonical

$$\frac{\text{tr} (OP_\delta(E))}{\text{tr} P_\delta(E)} \Rightarrow O(E)$$

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# energy filter

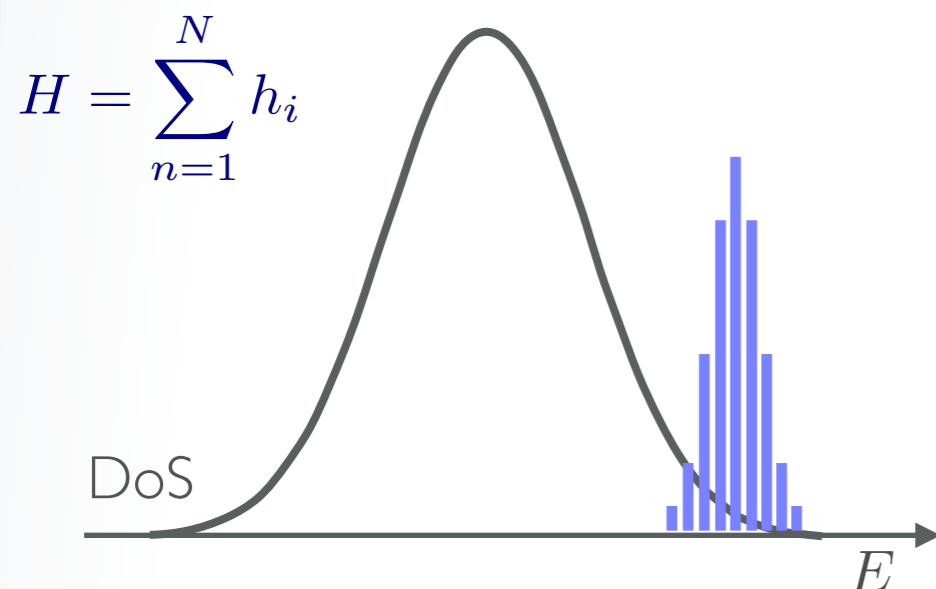
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equivalent to diagonal ensemble of a certain pure state



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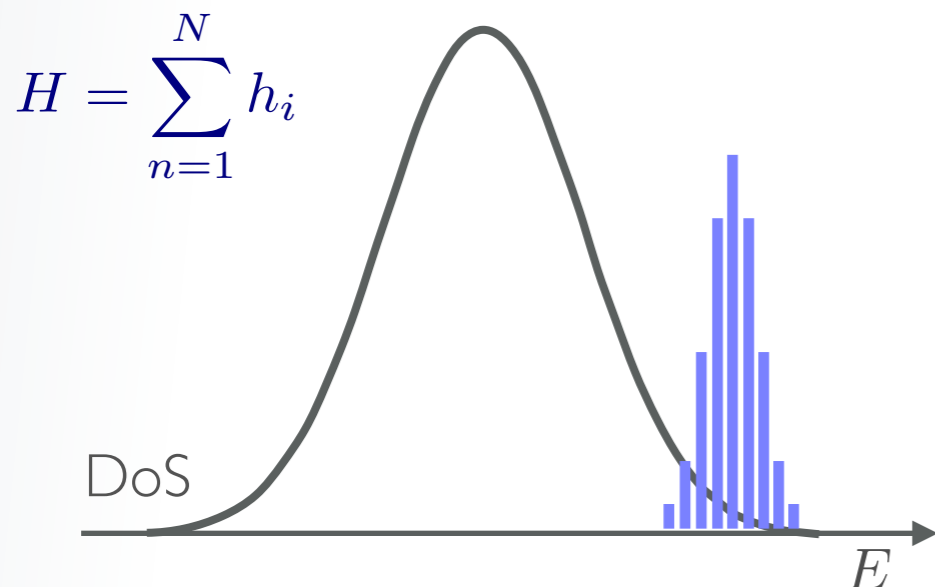
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equivalent to diagonal ensemble of a certain pure state

reached only after long time evolution



# computing observables

Gaussian filter  $\Rightarrow$  approximated by series of evolutions

$$\exp\left[-\frac{(H - E)^2}{2\delta^2}\right] \approx \sum_{m=-R}^R c_m e^{-i2mE/\alpha} e^{i2mH/\alpha}$$

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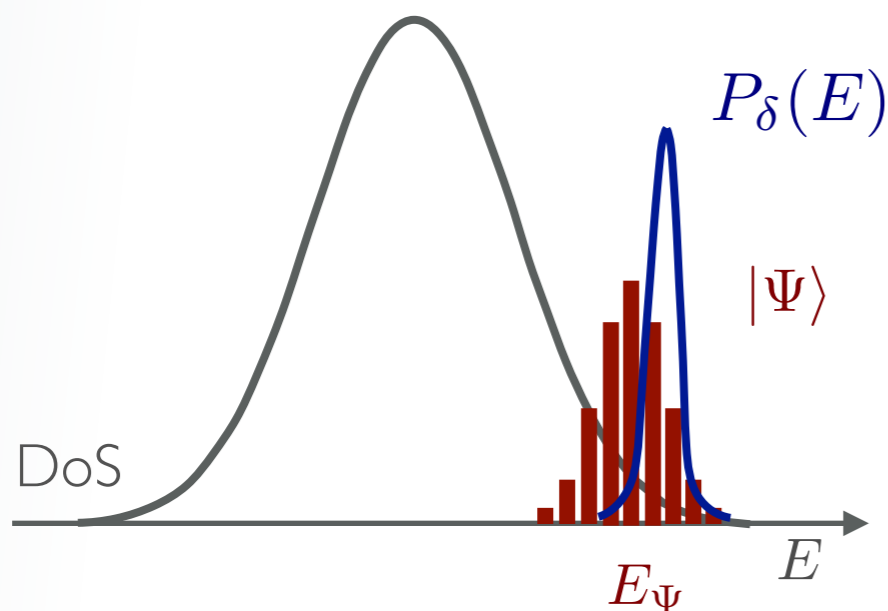
quantum inspired classical method

# computing observables

e.g. (broadened) local density of states

$$D_{\delta, \Psi}(E) = \langle \Psi | P_{\delta}(E) | \Psi \rangle$$

$$P_{\delta}(E) = \sum_{m=-R}^R c_m e^{-i(H-E)t_m} \quad \begin{aligned} R &= xN/\delta \\ t_m &= 2m/N \end{aligned}$$



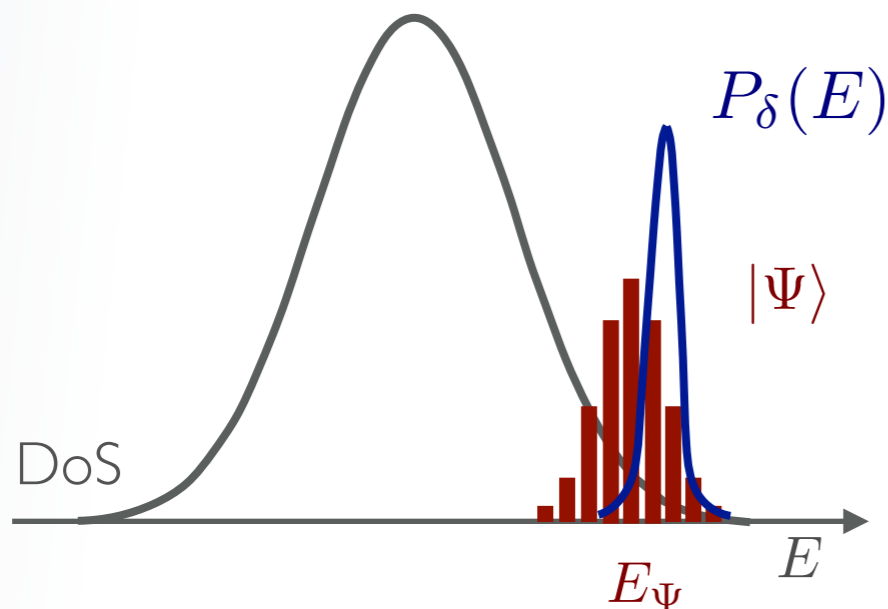
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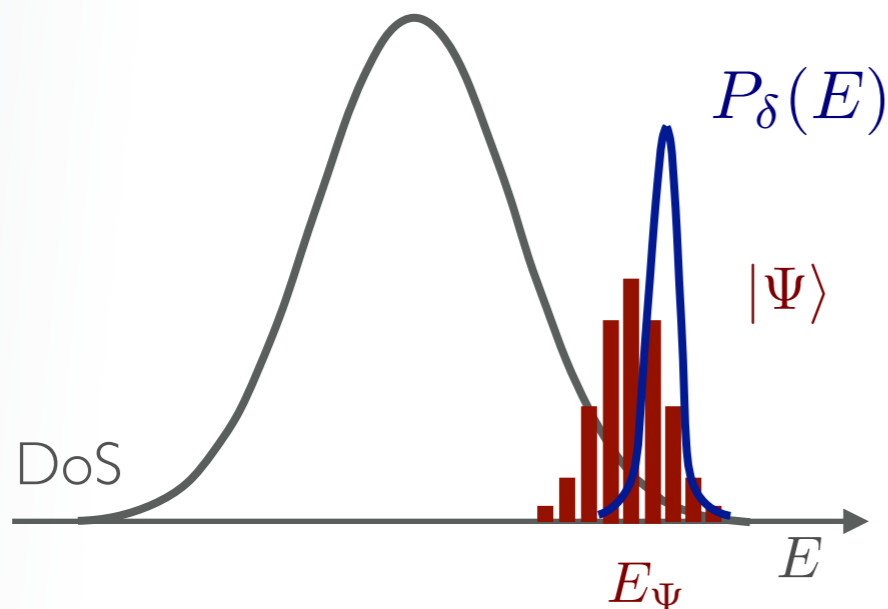
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Q simulator: prepare state,  
evolve, measure

$$t_{\max} \propto \frac{1}{\delta}$$

# computing observables

can compute observables:

$$\tilde{A}_{\delta, \Psi}(E) = \frac{\langle \Psi | P_{\delta}(E) A P_{\delta}(E) | \Psi \rangle}{\langle \Psi | P_{\delta}(E)^2 | \Psi \rangle}$$

potential problem: denominator too small

shown to be large enough in vicinity of  $E_{\Psi}$

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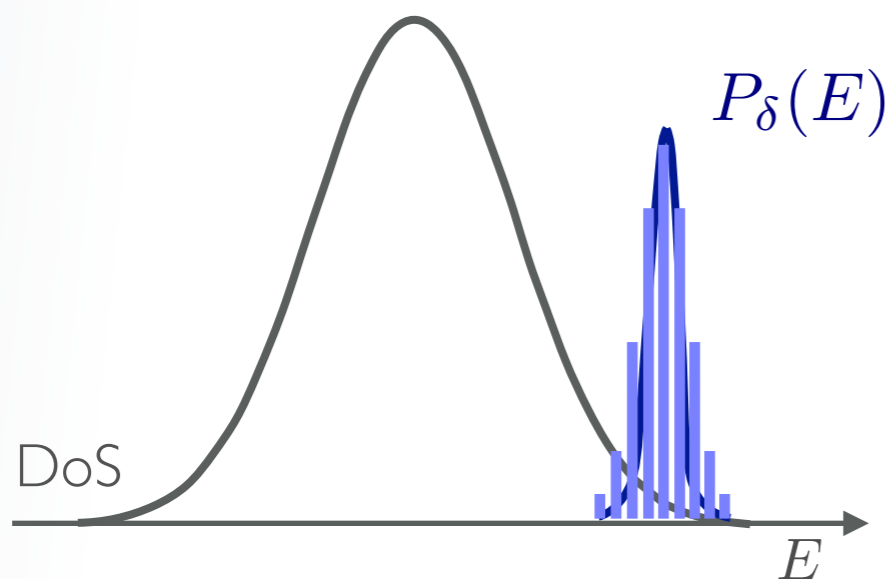
better convergence

$$A_{\delta}(E) = \frac{\text{tr}[A P_{\delta}(E)]}{\text{tr}[P_{\delta}(E)]}$$

# computing observables

expectation values in filter ensemble

$$A_\delta(E) = \frac{\text{tr}[AP_\delta(E)]}{\text{tr}[P_\delta(E)]}$$

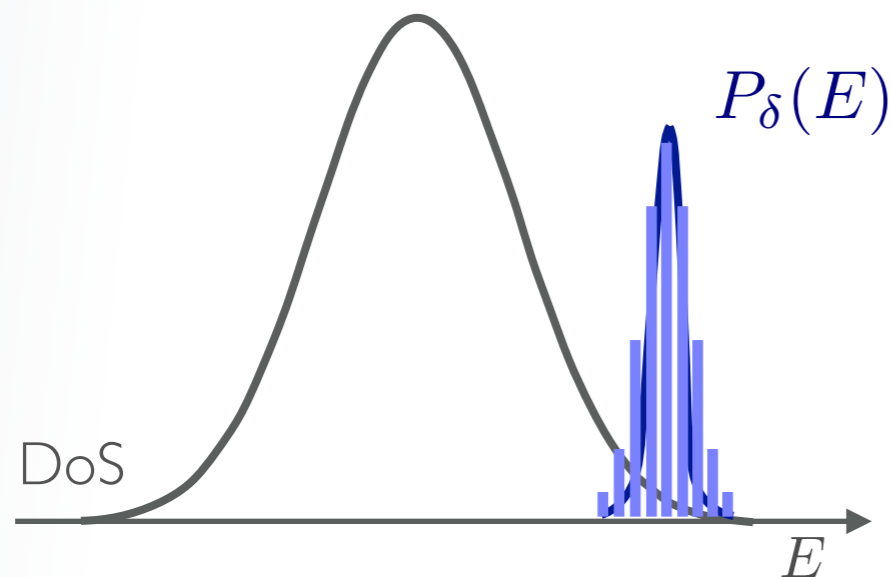




# algorithm

can be computed using Monte Carlo sampling

$$A_{\delta}(E) = \frac{\sum_{\Psi} \langle \Psi | A P_{\delta}(E) | \Psi \rangle}{\sum_{\Psi'} \langle \Psi' | P_{\delta}(E) | \Psi' \rangle}$$



# algorithm

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importance sampling (classical)

# algorithm

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computed by filter algorithm

importance sampling (classical)

algorithm

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computed by filter algorithm

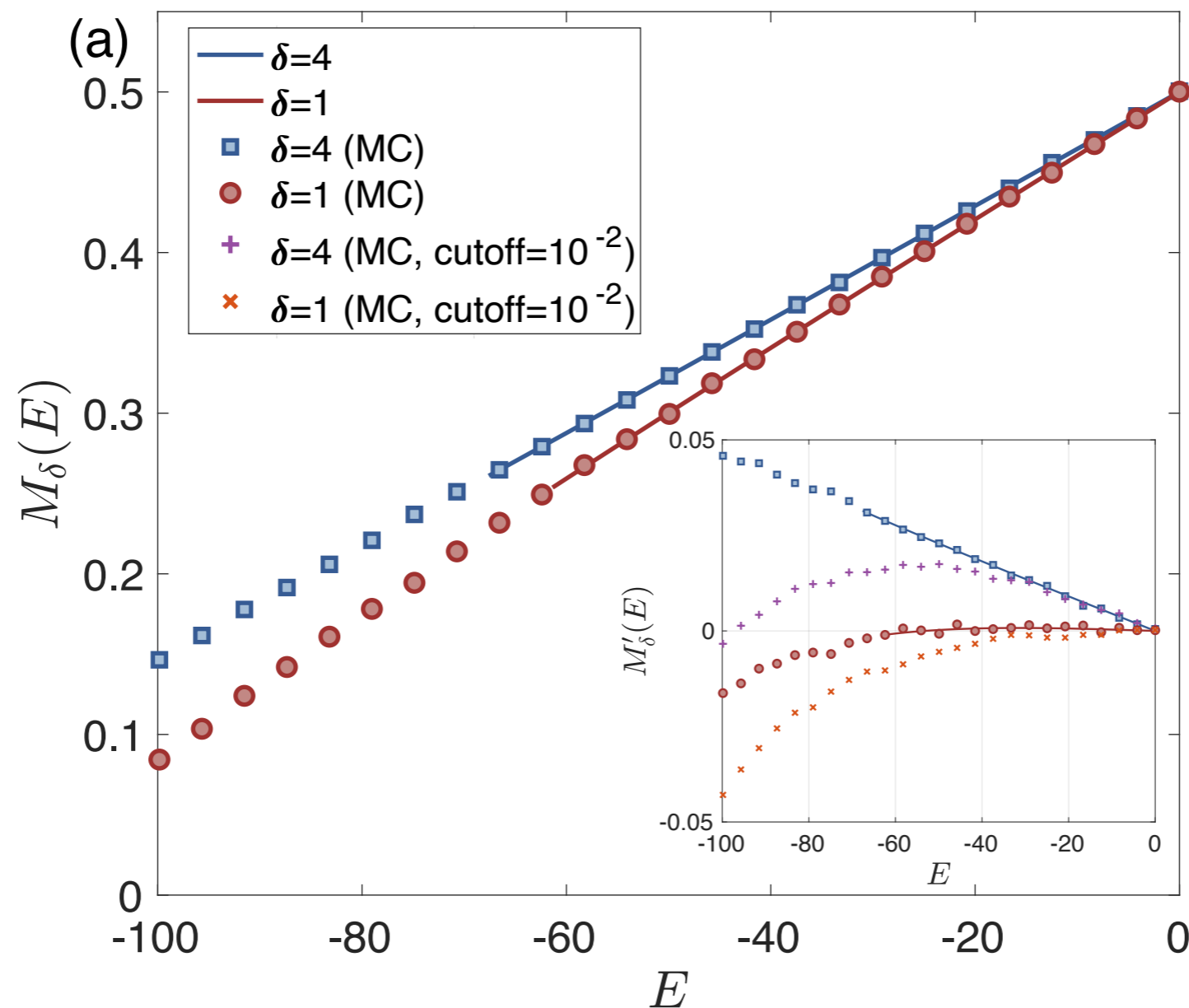
importance sampling (classical)

simulating classically with TNS we can reach  $\delta \sim 1/\log N$

# hybrid strategy for microcanonical quantities

Ising model,  $N=100$

$$M = \frac{1}{2N} \sum_{n=1}^N (\sigma_{n,z} + 1)$$



weak ETH probe: diagonal part

weak ETH probe: diagonal part

$$O_{\alpha\beta} = O(\bar{E})\delta_{\alpha\beta} + e^{-\frac{S(\bar{E})}{2}} f_O(\bar{E}, \omega) R_{\alpha\beta}$$

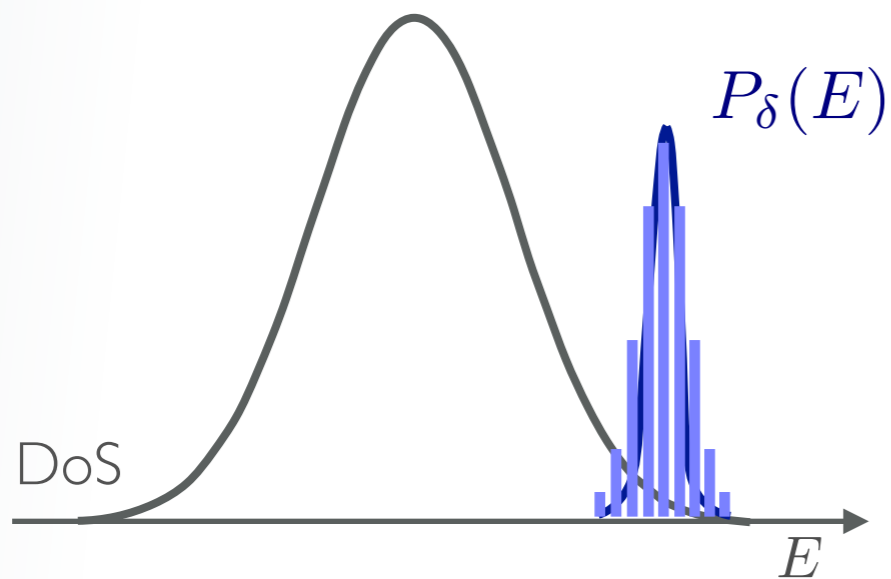
converge to thermal for large systems





# TNS simulation

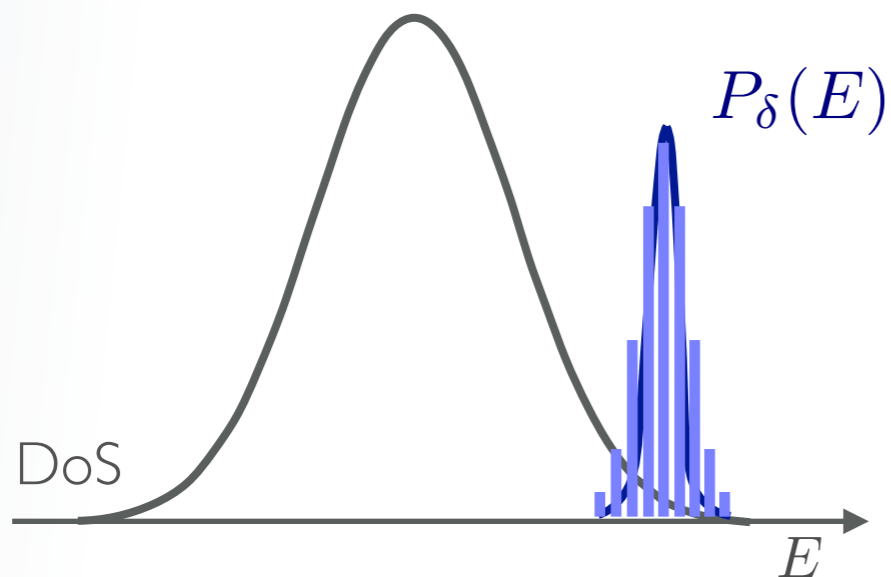
$$A_\delta(E) = \frac{\text{tr}[AP_\delta(E)]}{\text{tr}[P_\delta(E)]}$$



# TNS simulation

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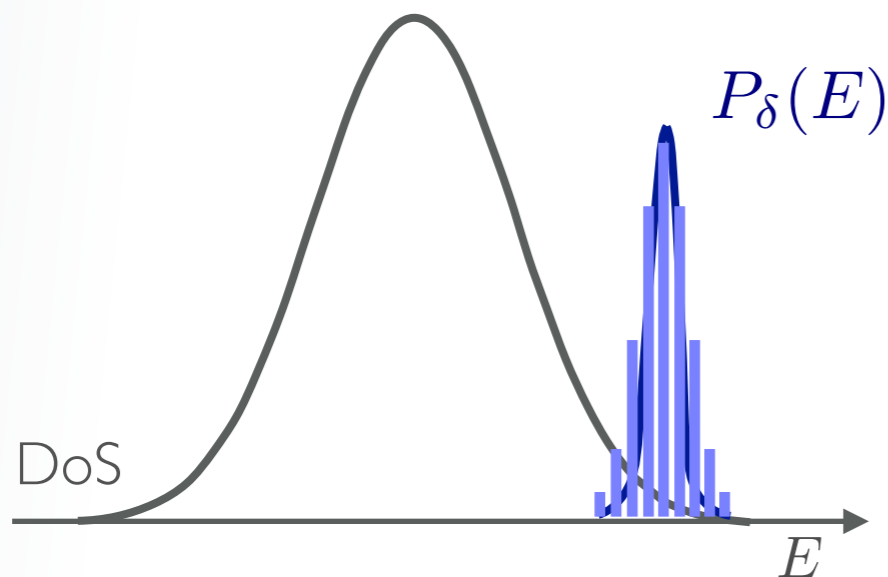
$$\text{ETH} \Rightarrow A_{nn} \sim A(E_n) + \delta \frac{dA}{dE}$$



# TNS simulation

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$$\text{ETH} \Rightarrow A_{nn} \sim A(E_n) + \delta \frac{dA}{dE}$$



for intensive quantities

$$\delta/N \rightarrow 0 \text{ as } N \rightarrow \infty$$

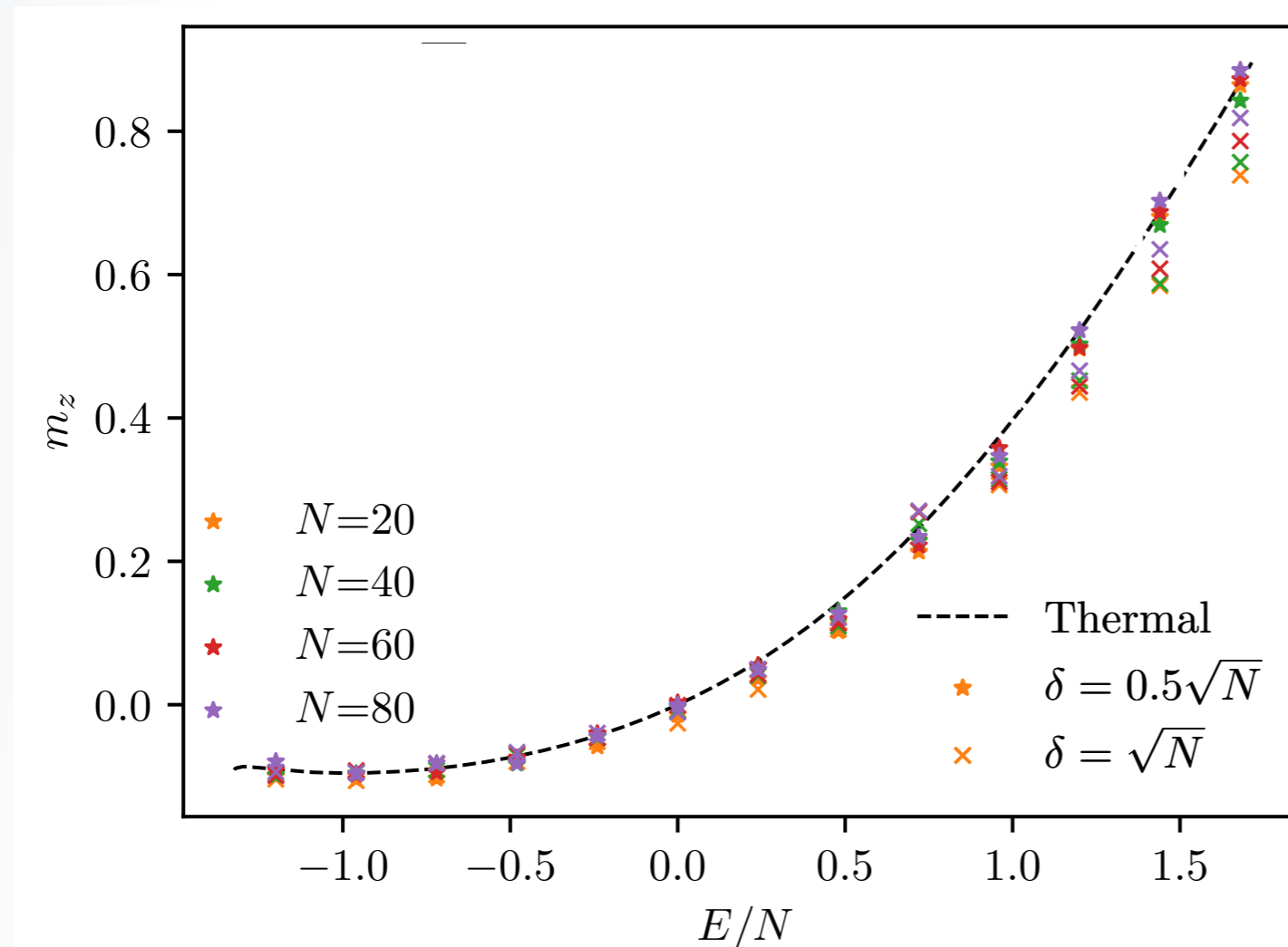
is enough for convergence to  
microcanonical

test on non-integrable Ising chain

$$H = -J \sum_{i=1}^{N-1} \sigma_i^z \sigma_{i+1}^z + \sum_{i=1}^N (g\sigma_i^x + h\sigma_i^z)$$

# TNS simulation

non-integrable quantum Ising chain



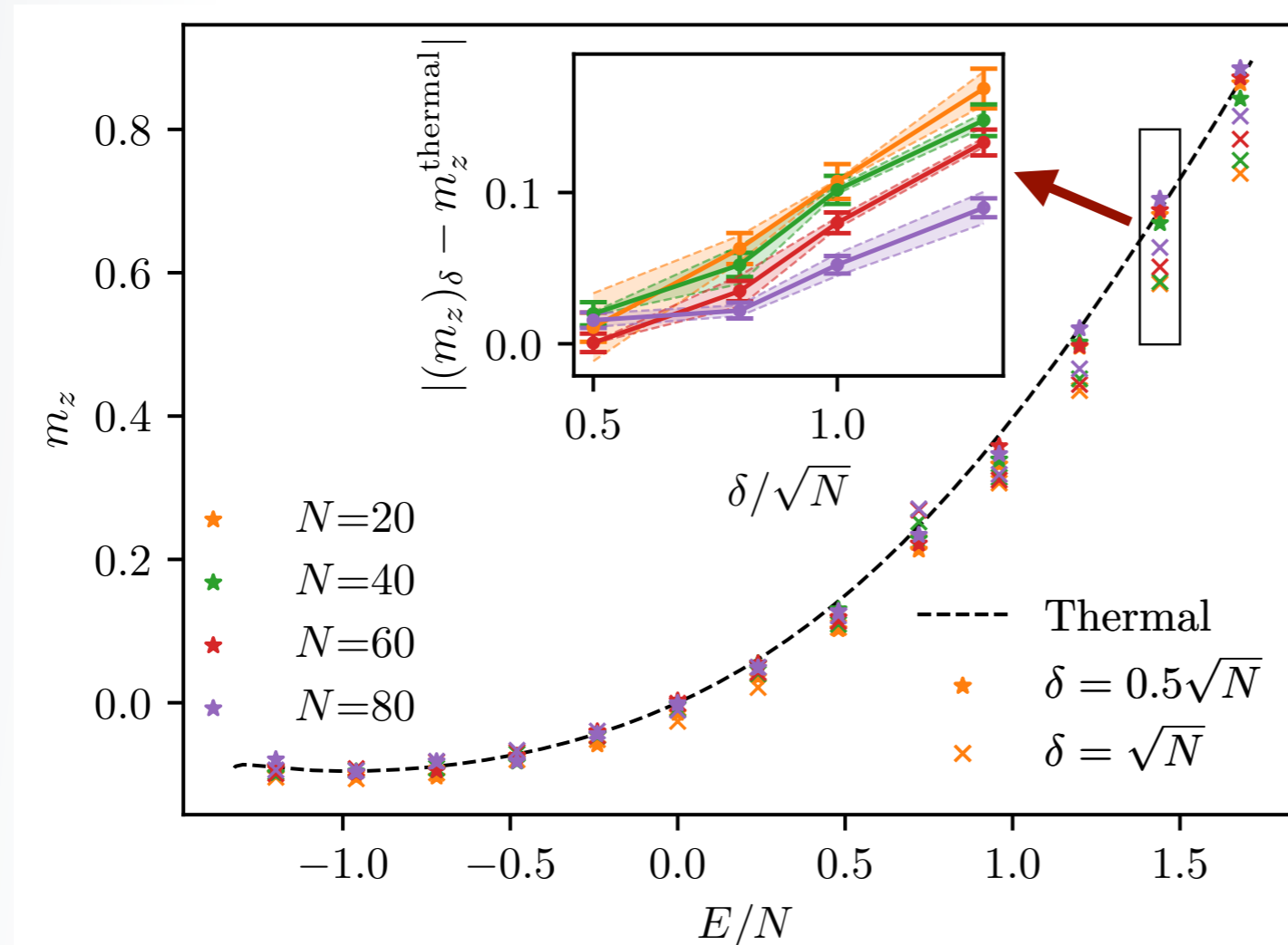
microcanonical properties

average magnetization

MPO + sampling over product states

# TNS simulation

non-integrable quantum Ising chain



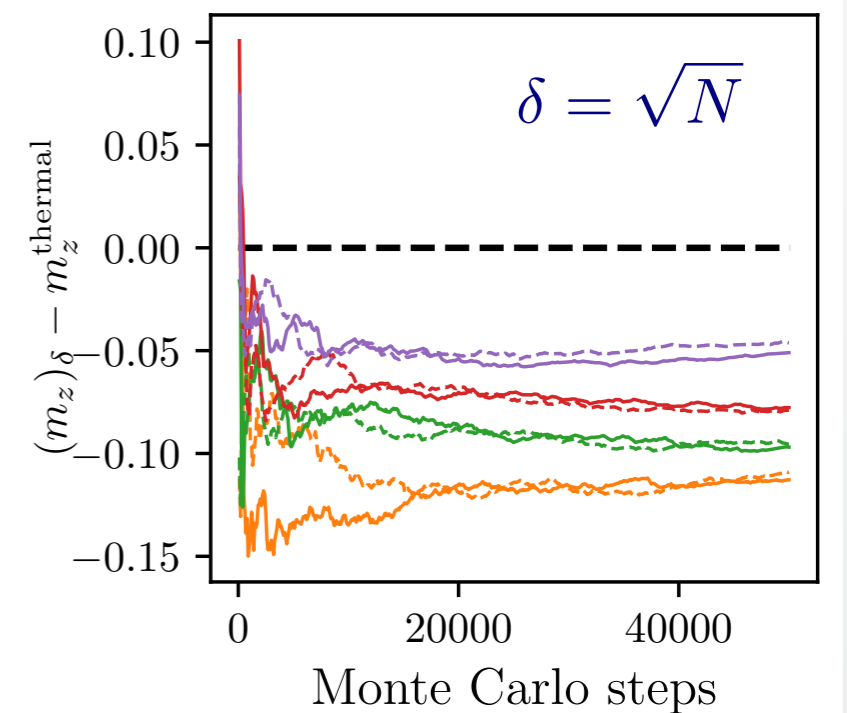
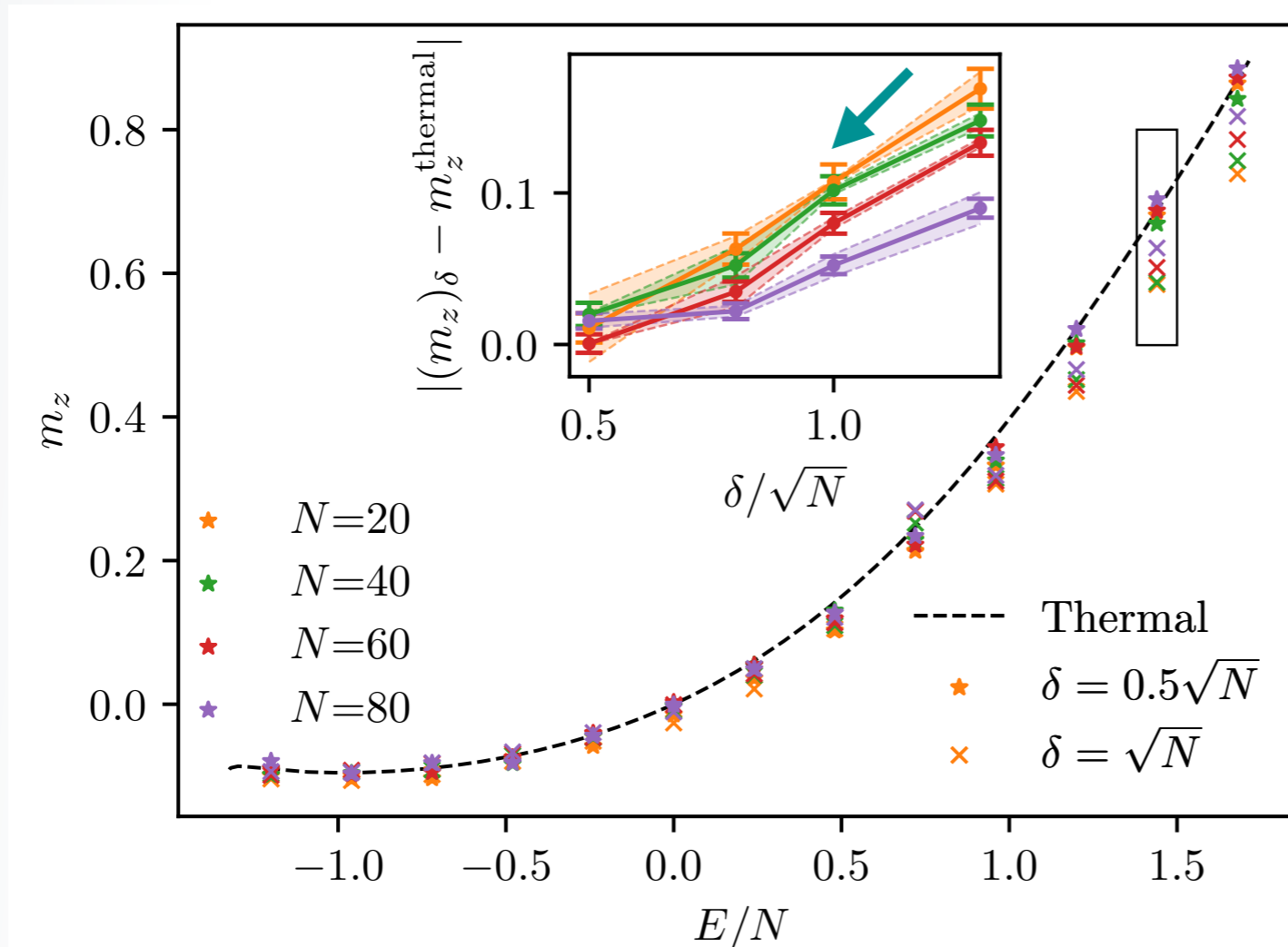
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# TNS simulation

non-integrable quantum Ising chain



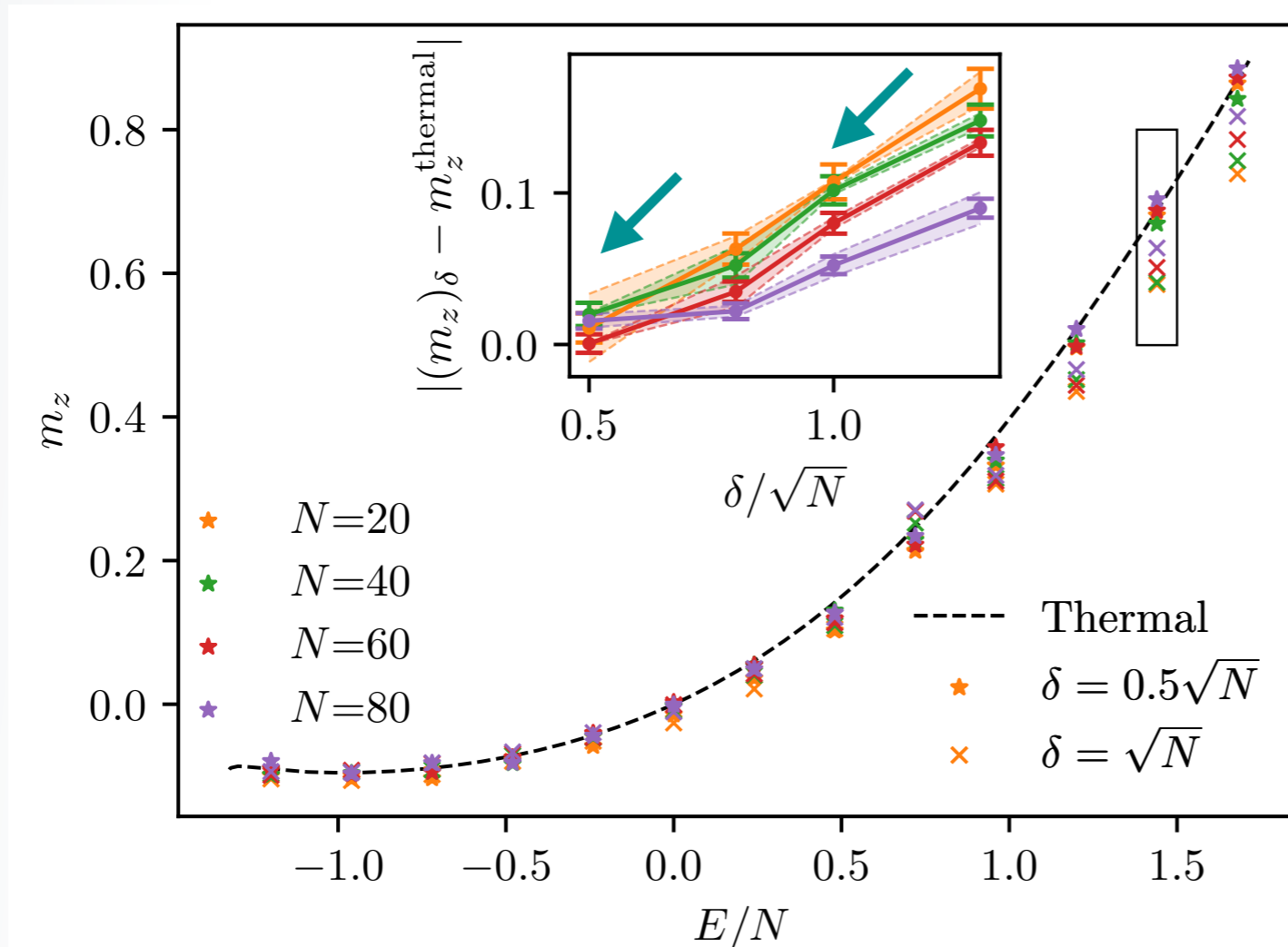
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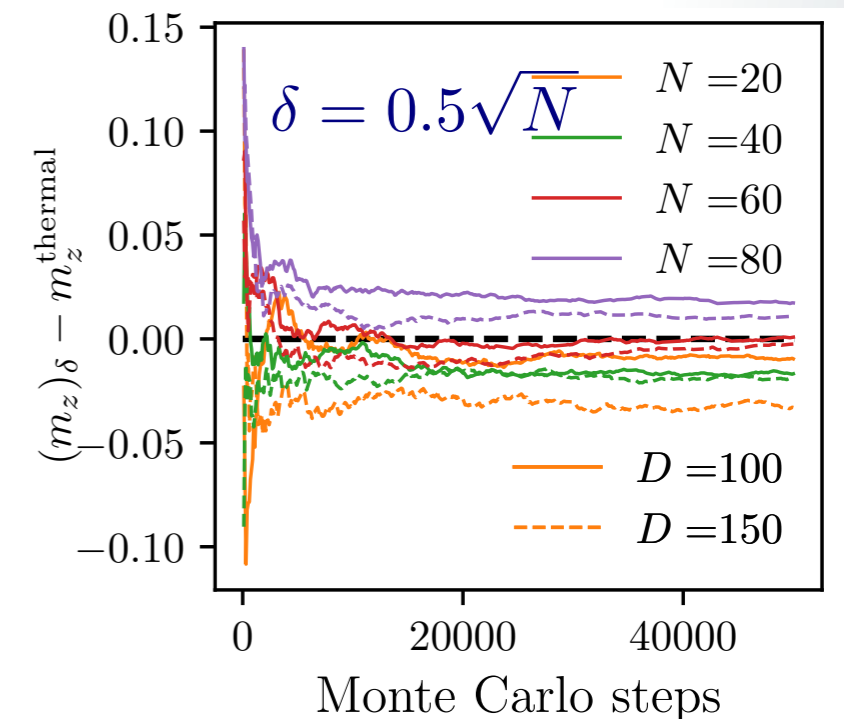
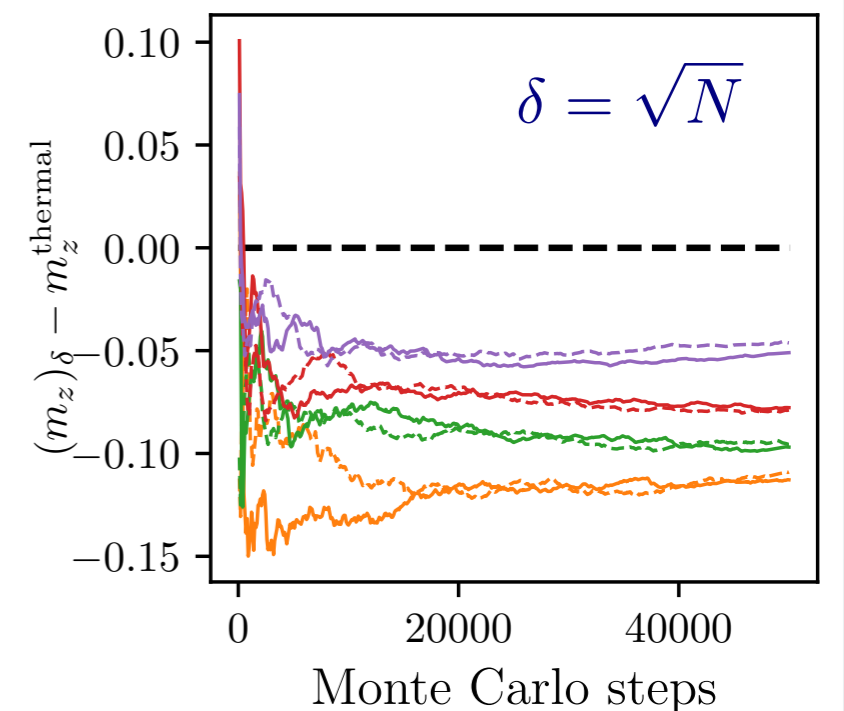
non-integrable quantum Ising chain



microcanonical properties

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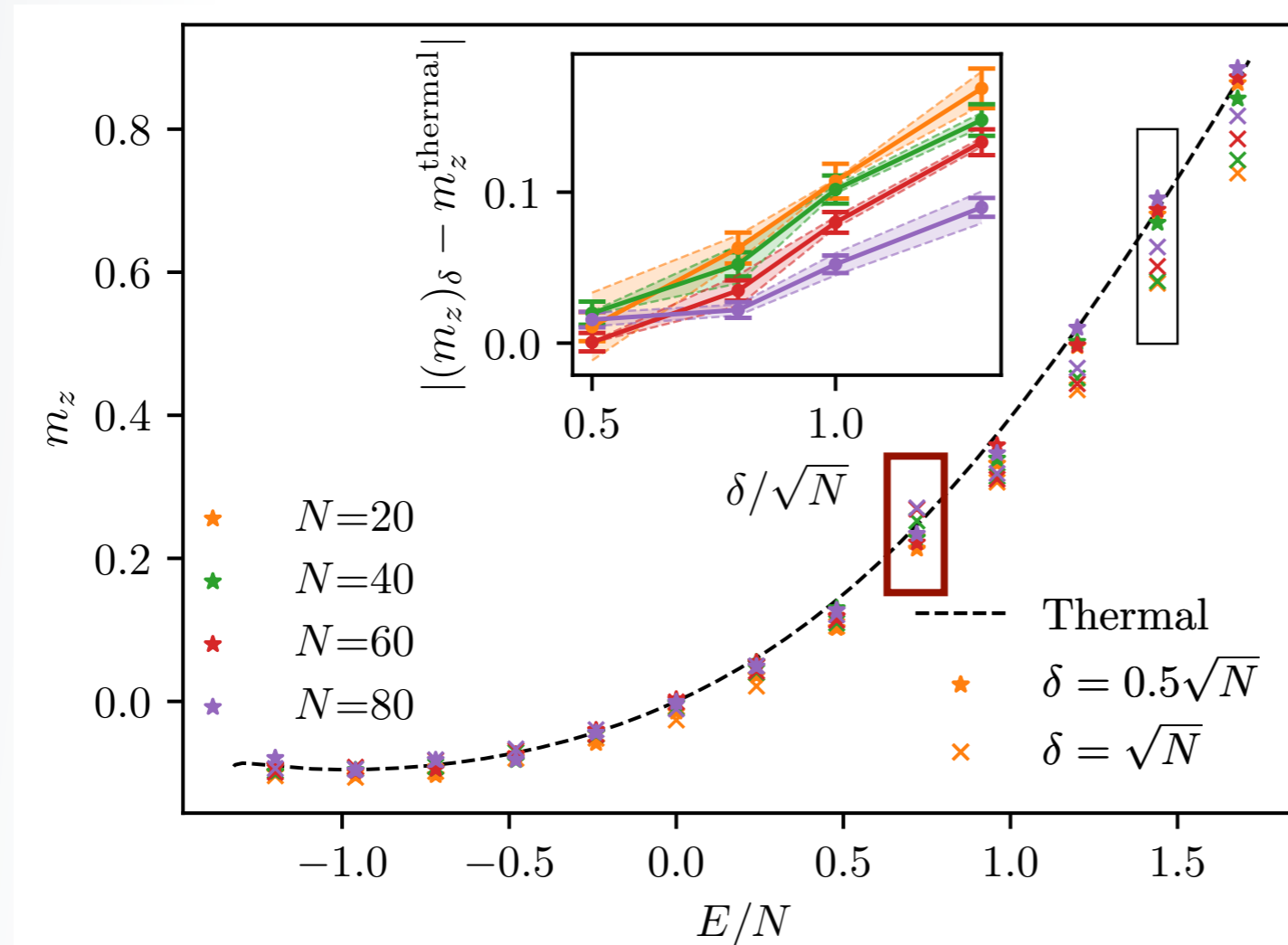
MPO + sampling over product states





# TNS simulation

non-integrable Ising model

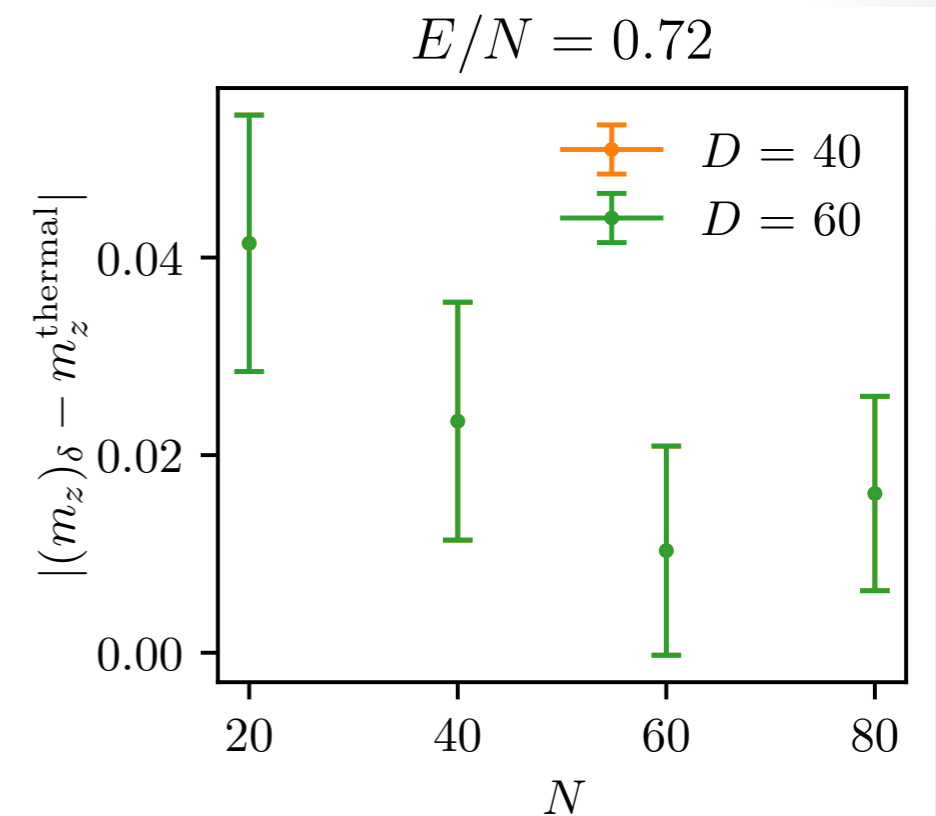
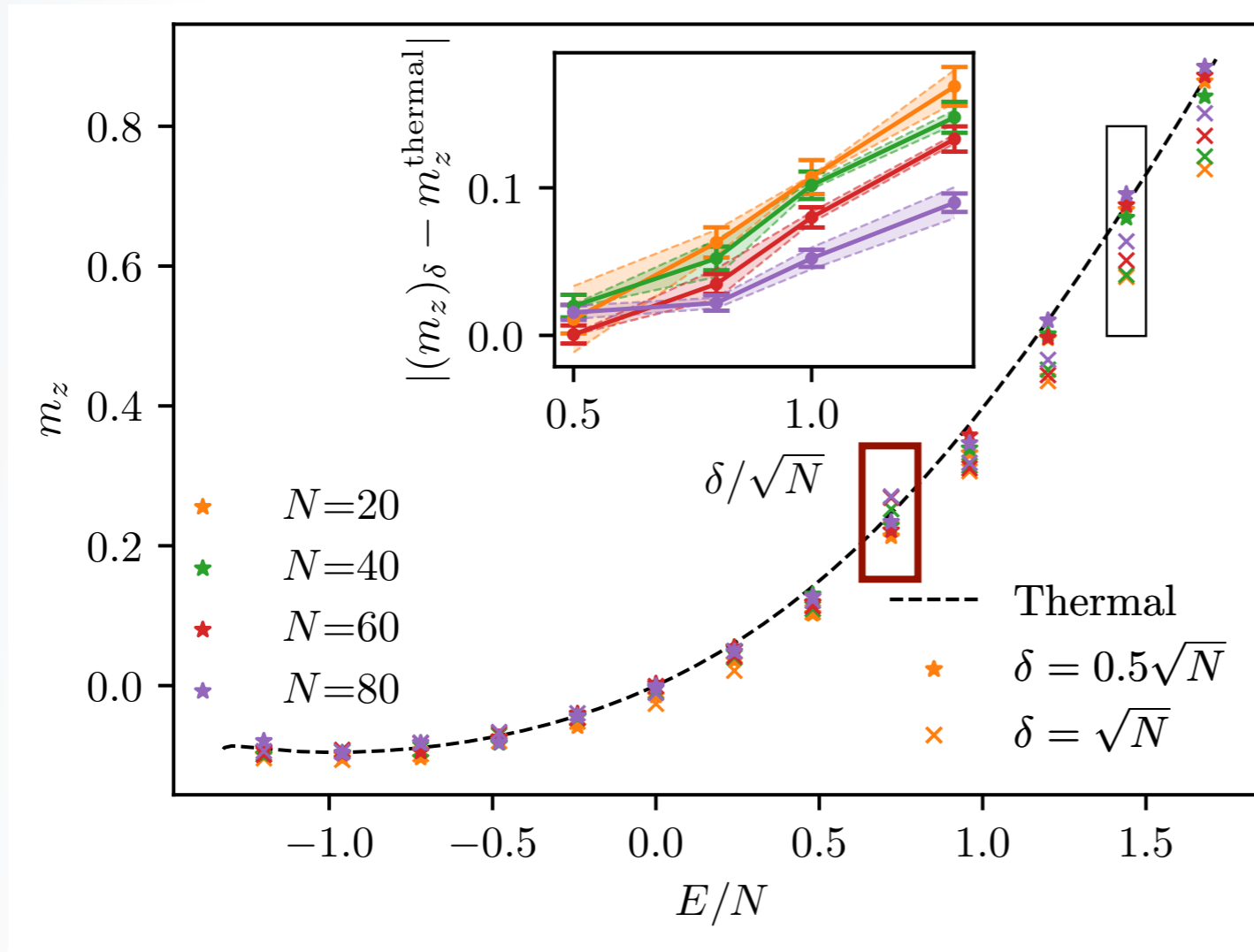


microcanonical properties  
average magnetization

# TNS simulation

non-integrable Ising model

also sampling over product states and evolve MPS



microcanonical properties  
average magnetization

more challenging: off-diagonal part of ETH

more challenging: off-diagonal part of ETH

$$O_{\alpha\beta} = O(\bar{E})\delta_{\alpha\beta} + e^{-\frac{S(\bar{E})}{2}} f_O(\bar{E}, \omega) R_{\alpha\beta}$$

structure function



Luitz, Bar Lev, PRL2016; Mondaini, Rigol 2017; Brenes et al PRL2020, PRB 2020...;  
Schönle et al PRB2021; Essler, de Klerk, 2307.12410; ....

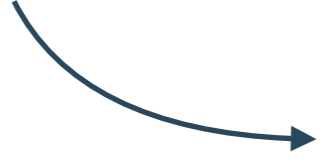
filter ensemble as ETH probe

function  $f_O(\bar{E}, \omega)$  related to two-time correlators

## filter ensemble as ETH probe

function  $f_O(\bar{E}, \omega)$  related to two-time correlators

$$C_O(t) = \text{tr}(\rho_E O(t) O(0))$$


$$S_O(\omega) = \sum_{\alpha\beta} \rho_{\alpha\alpha} |O_{\alpha\beta}|^2 \delta(\omega - E_\beta + E_\alpha)$$

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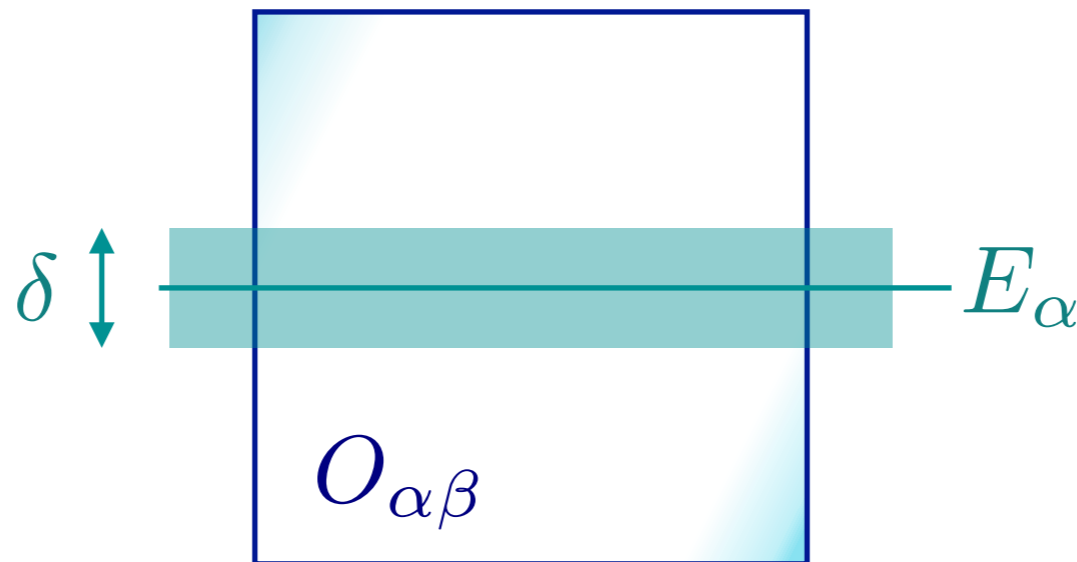
# filter ensemble as ETH probe

function  $f_O(\bar{E}, \omega)$  related to two-time correlators

$$C_O(t) = \text{tr}(\rho_E O(t) O(0))$$

for filter ensemble

$$S_O(\omega) = \sum_{\alpha\beta} \rho_{\alpha\alpha} |O_{\alpha\beta}|^2 \delta(\omega - E_\beta + E_\alpha)$$



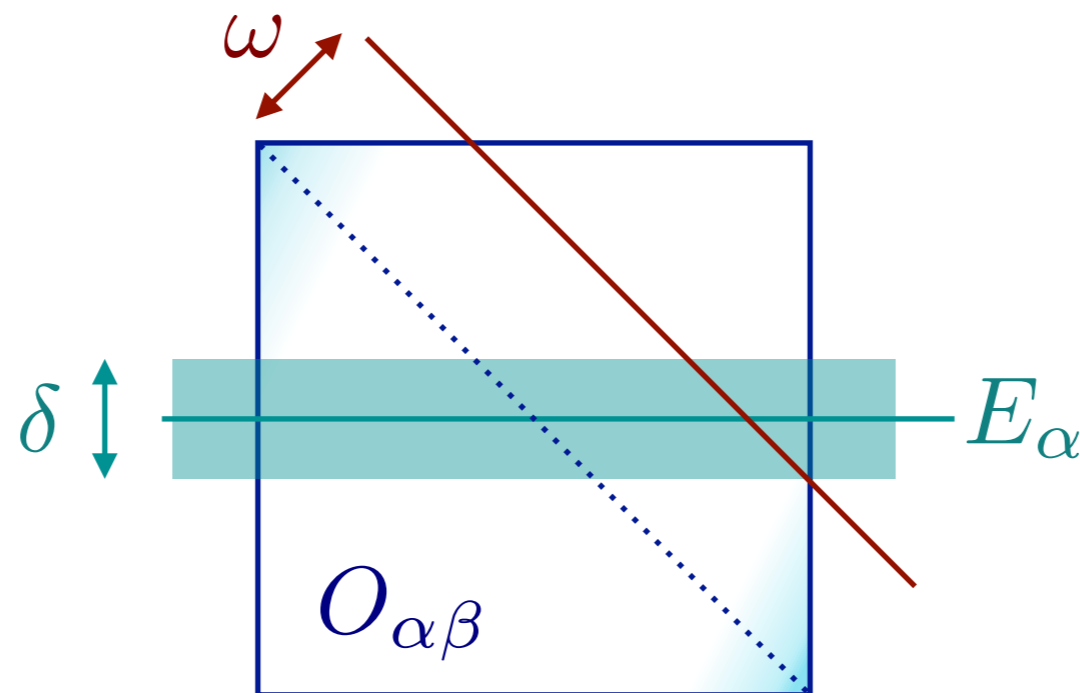


# filter ensemble as ETH probe

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$$C_O(t) = \text{tr}(\rho_E O(t) O(0)) \quad \text{for filter ensemble}$$

$$S_O(\omega) = \sum_{\alpha\beta} \rho_{\alpha\alpha} |O_{\alpha\beta}|^2 \delta(\omega - E_\beta + E_\alpha)$$

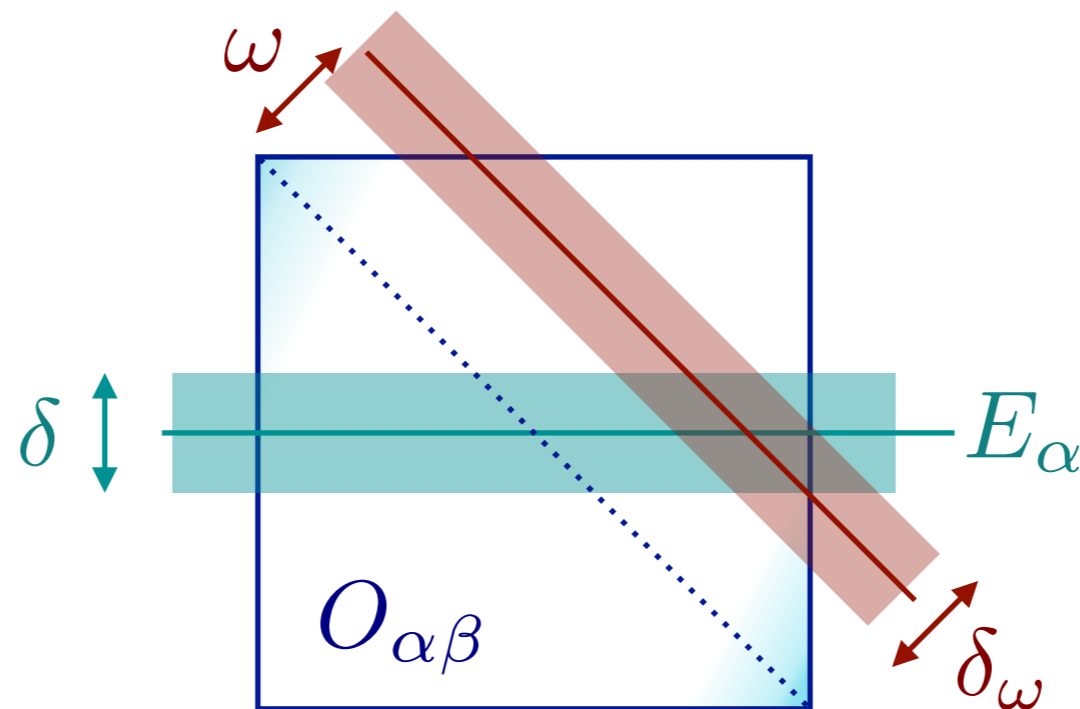


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$P_{\delta_\omega}(\omega)$   
filter in energy  
difference  
(commutator)

# filter ensemble as ETH probe

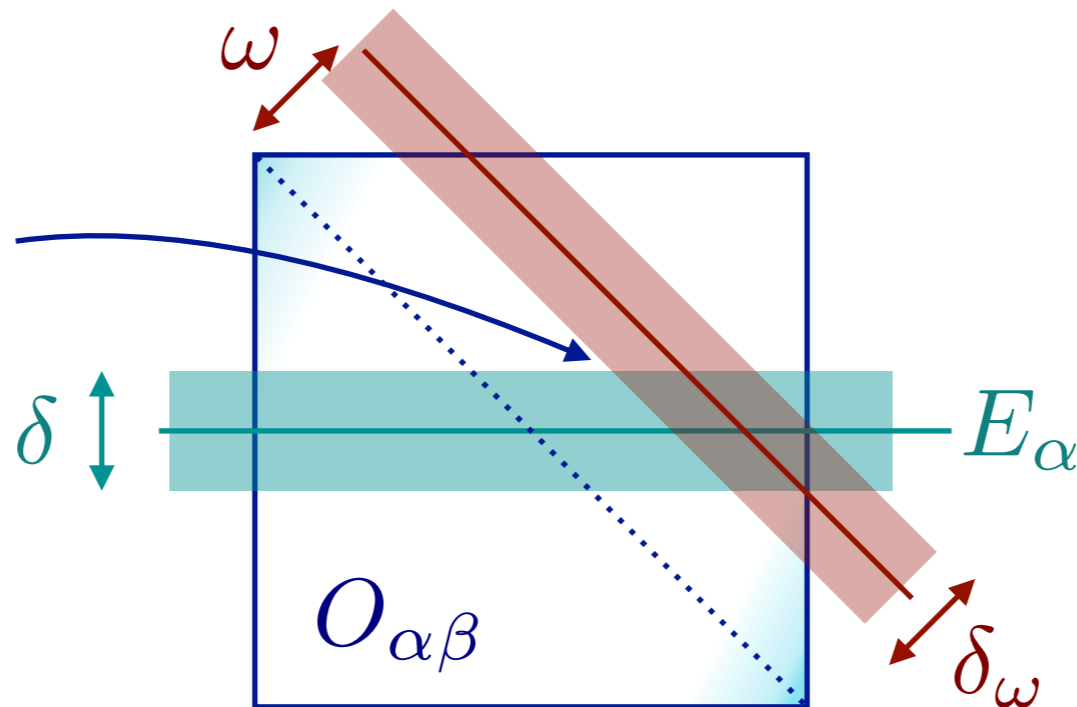
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$$S_O(\omega) = \sum_{\alpha\beta} \rho_{\alpha\alpha} |O_{\alpha\beta}|^2 \delta(\omega - E_\beta + E_\alpha)$$

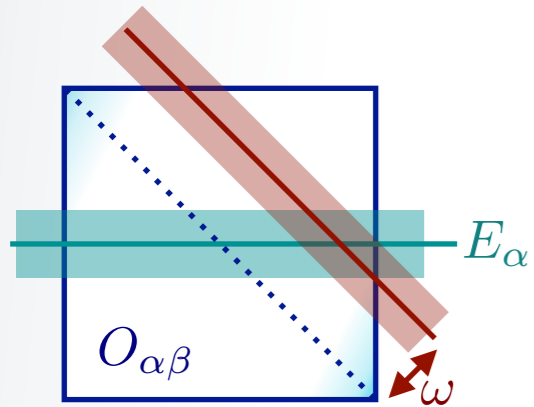
$$e^{S(E+\omega)} |O_{E,E+\omega}|^2$$

average times  
density of states  
factor



$P_{\delta_\omega}(\omega)$   
filter in energy  
difference  
(commutator)

# filter ensemble as ETH probe

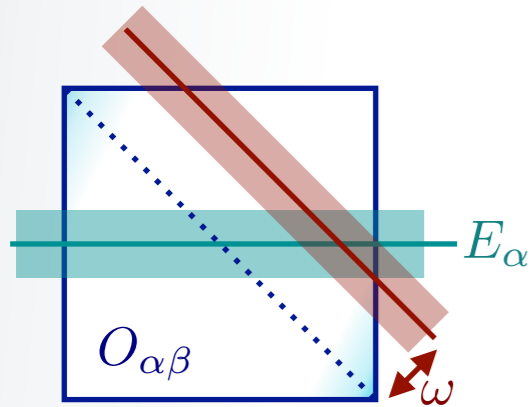


broadened spectral function of filter ensemble

$$S_O^\rho(\omega) \approx e^{S(E+\omega)} |O_{E,E+\omega}|^2$$

see also Pappalardi, Foini,  
Kurchan, 2304.10948

# filter ensemble as ETH probe



broadened spectral function of filter ensemble

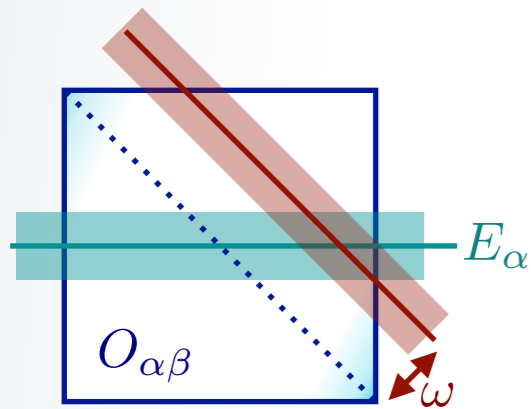
$$S_O^\rho(\omega) \approx e^{S(E+\omega)} |O_{E,E+\omega}|^2$$

see also Pappalardi, Foini,  
Kurchan, 2304.10948

$$\approx e^{\frac{S(E+\omega) - S(E)}{2}} |f_O(E + \omega/2, \omega)|^2$$

using ETH

# filter ensemble as ETH probe



broadened spectral function of filter ensemble

$$S_O^\rho(\omega) \approx e^{S(E+\omega)} |O_{E,E+\omega}|^2$$

see also Pappalardi, Foini,  
Kurchan, 2304.10948

$$\approx e^{\frac{S(E+\omega) - S(E)}{2}} |f_O(E + \omega/2, \omega)|^2$$

using ETH

entropy factor extracted from DoS  
calculation or eliminated from  $S_O^\rho(-\omega)$

# filter ensemble as ETH probe

errors from two filters with independent widths

filter ensemble

$$S_O^{P_\delta}(\omega) = e^{\frac{S(E+\omega) - S(E)}{2}} |f_O(E + \omega/2, \omega)|^2 \left[ 1 + O\left(\frac{\delta^2}{N^2}\right) \right]$$

# filter ensemble as ETH probe

errors from two filters with independent widths

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$$S_O^{P_\delta}(\omega) = e^{\frac{S(E+\omega) - S(E)}{2}} |f_O(E + \omega/2, \omega)|^2 \left[ 1 + O\left(\frac{\delta^2}{N^2}\right) \right]$$

filter in energy difference

$$S'^{P_\delta}(\omega) \sim S_O^{P_\delta}(\omega) + \frac{1}{2} \delta_\omega^2 \partial_\omega^2 S_O^{P_\delta}$$



# numerical results

model: quantum Ising chain

$$H = -J \sum_{i=1}^{N-1} \sigma_i^z \sigma_{i+1}^z$$

# numerical results

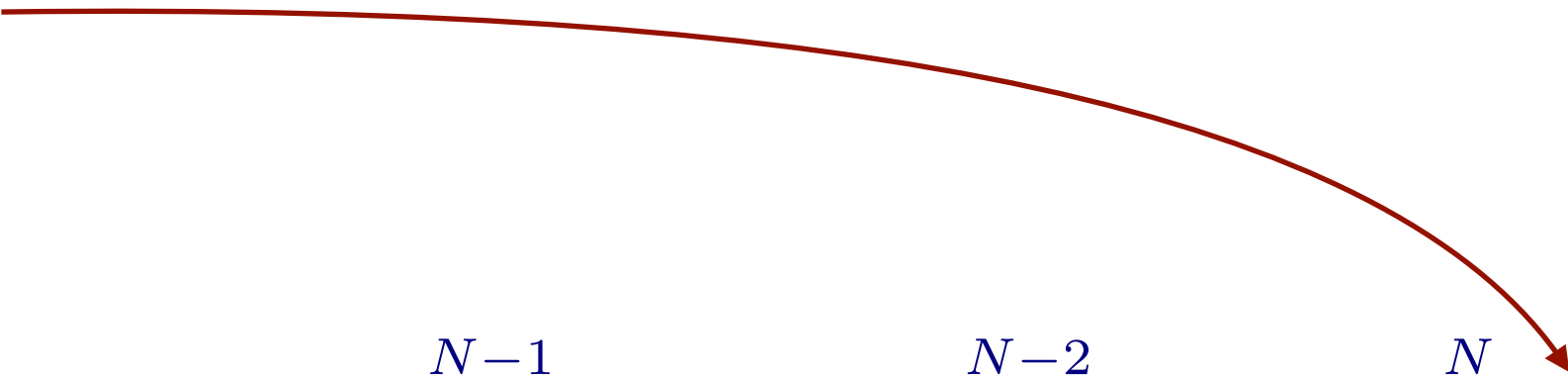
model: quantum Ising chain

$$H = -J \sum_{i=1}^{N-1} \sigma_i^z \sigma_{i+1}^z - J_2 \sum_{i=1}^{N-2} \sigma_i^z \sigma_{i+2}^z - \sum_{i=1}^N (g + r_i) \sigma_i^x$$

# numerical results

model: quantum Ising chain

integrable


$$H = -J \sum_{i=1}^{N-1} \sigma_i^z \sigma_{i+1}^z - J_2 \sum_{i=1}^{N-2} \sigma_i^z \sigma_{i+2}^z - \sum_{i=1}^N (g + r_i) \sigma_i^x$$

# numerical results

model: quantum Ising chain

integrable

non-  
integrable

$$H = -J \sum_{i=1}^{N-1} \sigma_i^z \sigma_{i+1}^z - J_2 \sum_{i=1}^{N-2} \sigma_i^z \sigma_{i+2}^z - \sum_{i=1}^N (g + r_i) \sigma_i^x$$

# numerical results

model: quantum Ising chain

integrable

non-  
integrable

disordered

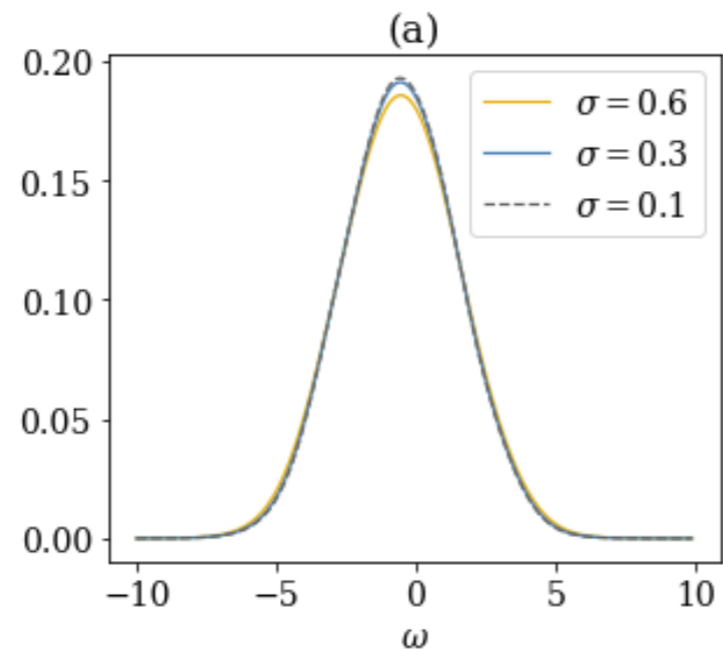
$$H = -J \sum_{i=1}^{N-1} \sigma_i^z \sigma_{i+1}^z - J_2 \sum_{i=1}^{N-2} \sigma_i^z \sigma_{i+2}^z - \sum_{i=1}^N (g + r_i) \sigma_i^x$$

# broadened spectral function

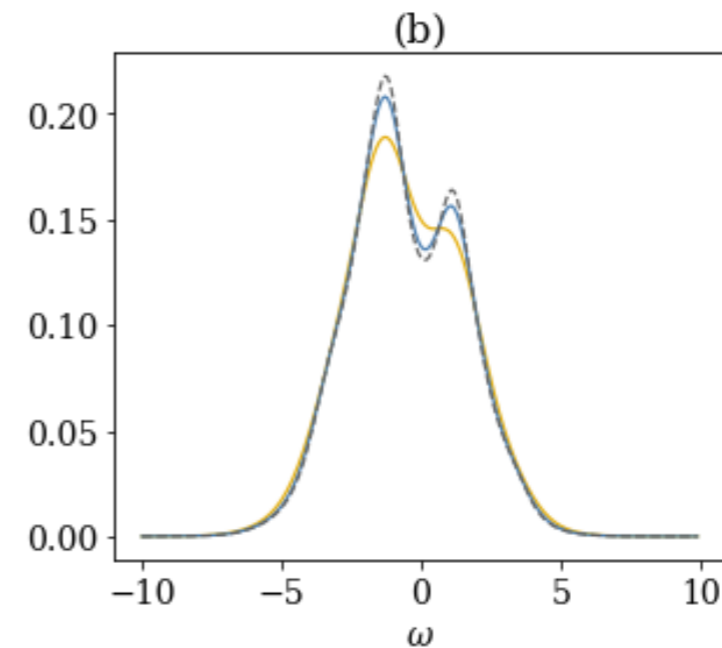
$$N = 40$$

$$E/N = 0.5$$

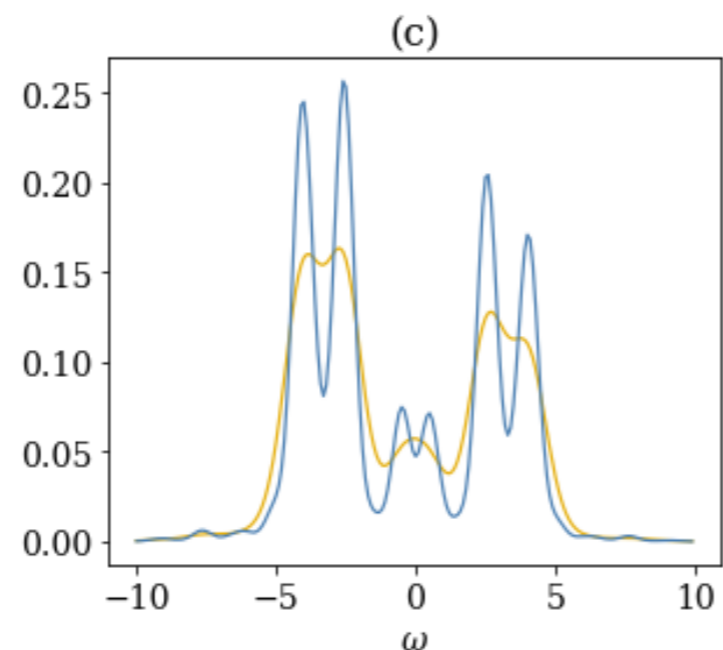
integrable



non-integrable



disordered

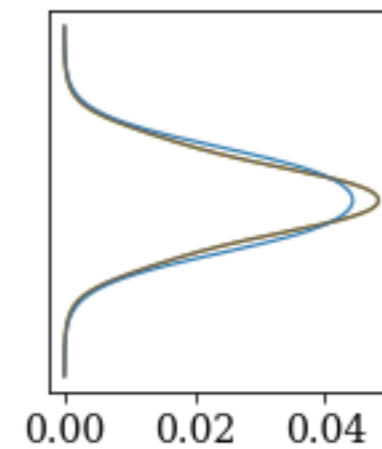
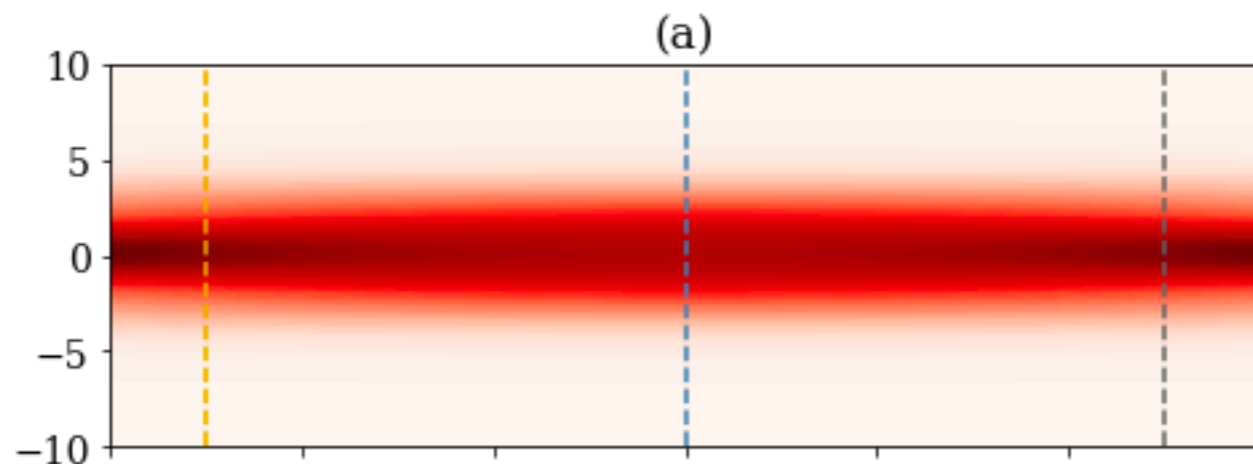


sufficiently narrow filters

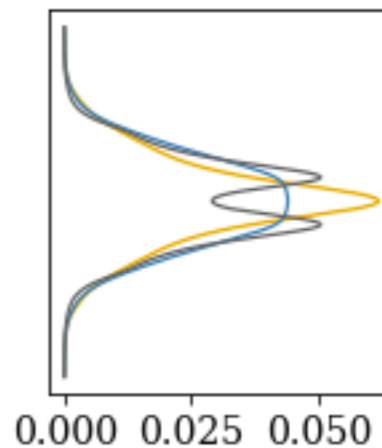
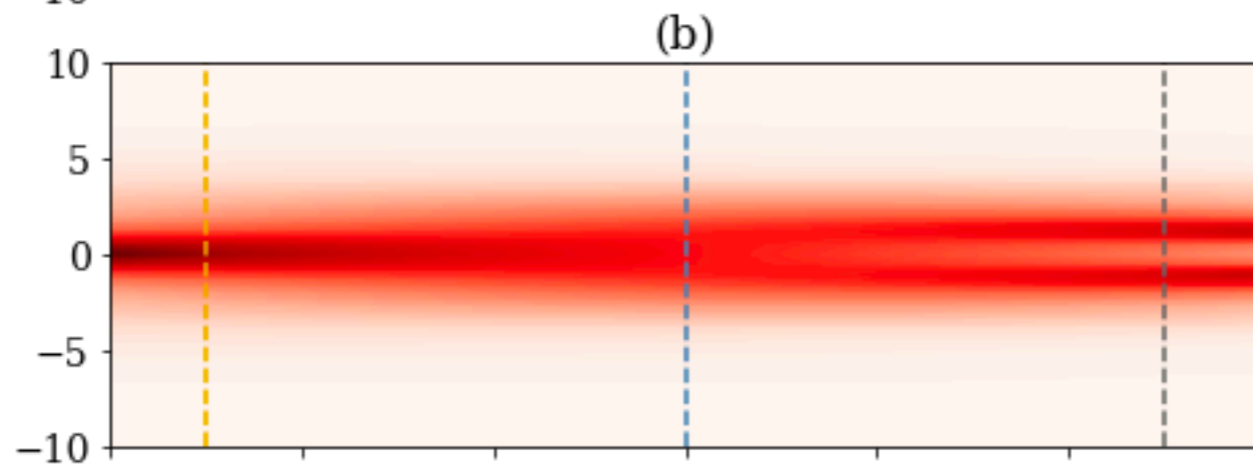
mapping out the function  $|f_O(E, \omega)|^2$

$N = 40$

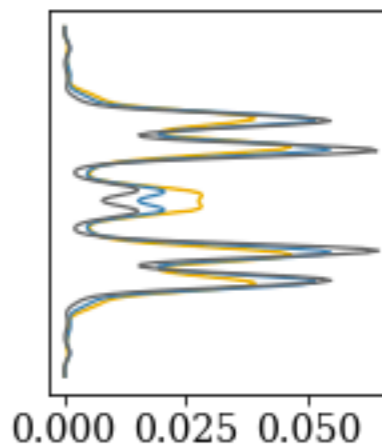
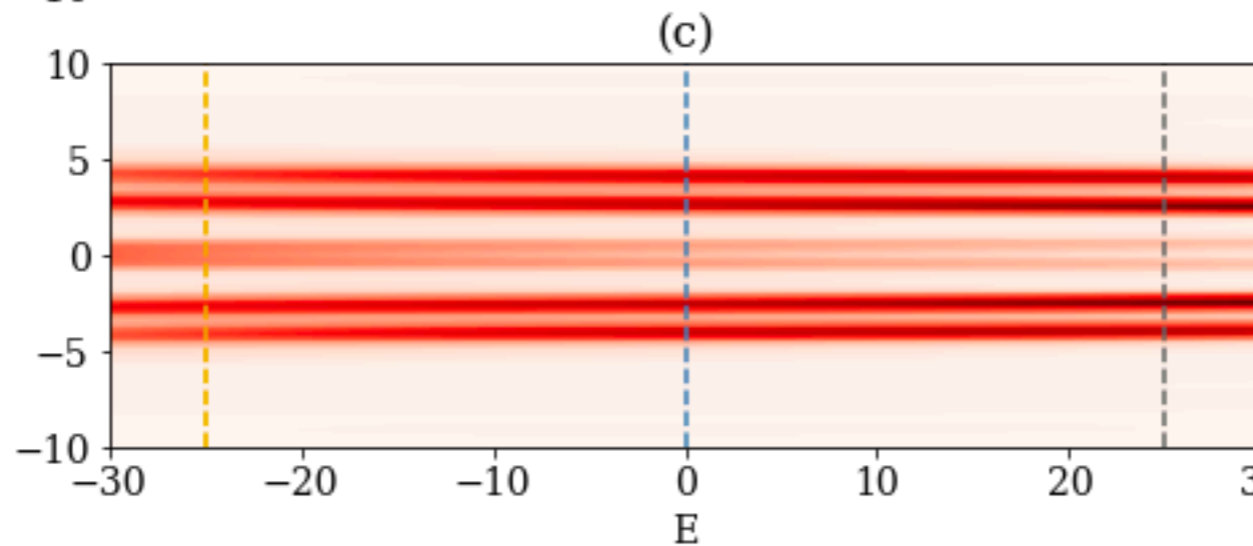
integrable



non-  
integrable



disordered

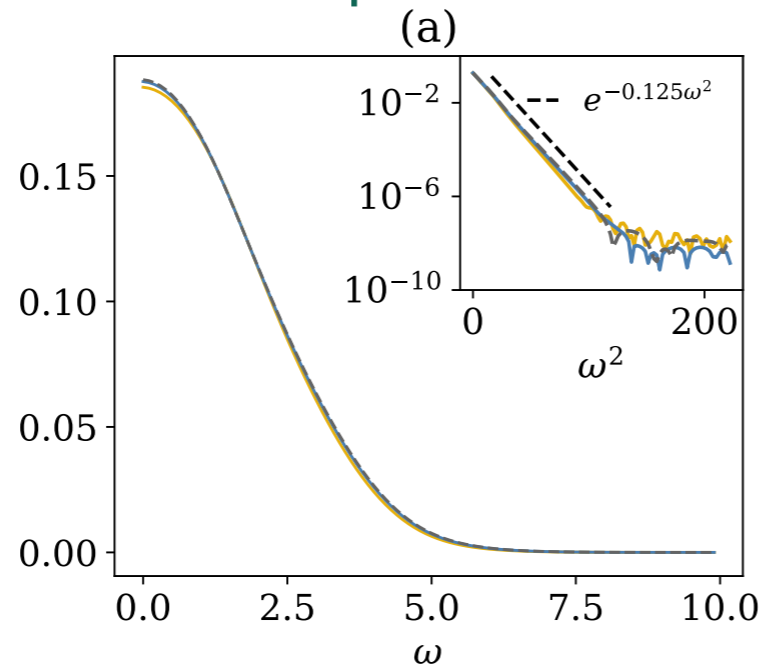


# $\omega$ -dependence of $|f_O(E, \omega)|^2$

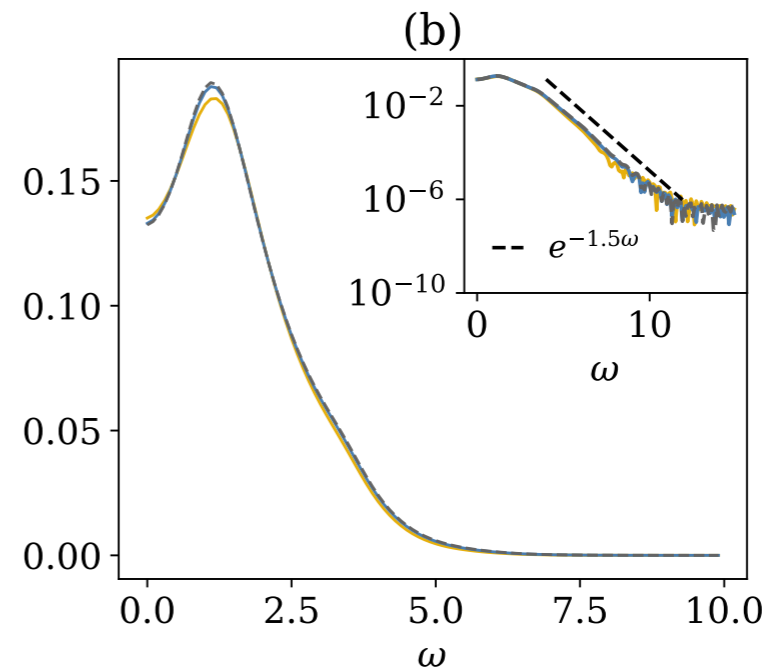
$$N = 20 - 60$$

$$E/N = 0.5$$

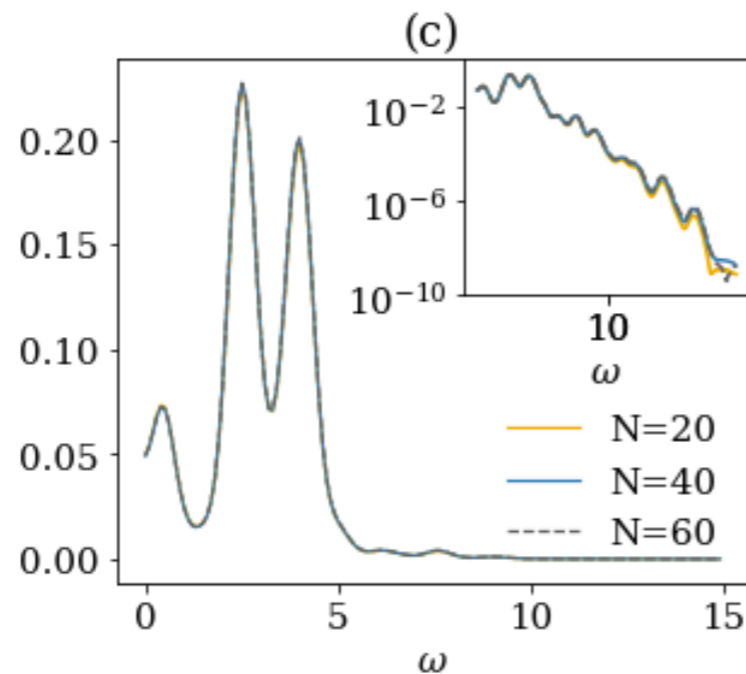
integrable



non-integrable



disordered



converged in size  
asymptotic behaviour



# other results with TNS + filter method

diagonal ensemble

Ising non-integrable

$$H_{\text{Ising}} = J \sum_i \sigma_z^{[i]} \sigma_z^{[i+1]} + g \sum_i \sigma_x^{[i]} + h \sum_i \sigma_z^{[i]}$$

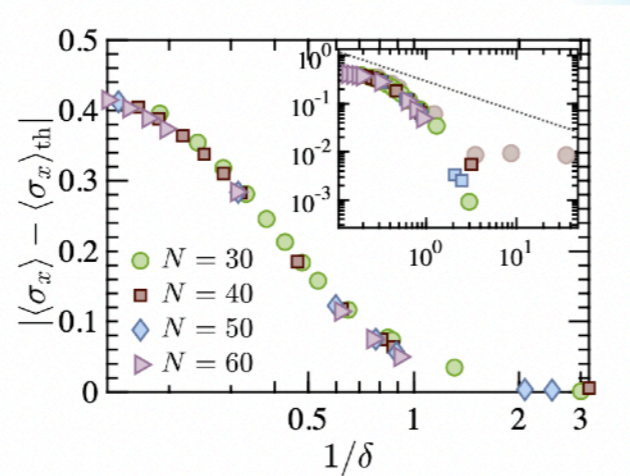
$$J = 1, g = -1.05, h = 0.5$$

product initial states

$$|Z+\rangle := |0\rangle^{\otimes N}$$

Çakan et al., PRB103, 115113 (2021)

local observable



→ approach thermal as filter narrows

entropy in diagonal ensemble  
grows  $\sim$ linearly with  $N$

# other results with TNS + filter method

## diagonal ensemble

### Ising non-integrable

$$H_{\text{Ising}} = J \sum_i \sigma_z^{[i]} \sigma_z^{[i+1]} + g \sum_i \sigma_x^{[i]} + h \sum_i \sigma_z^{[i]}$$

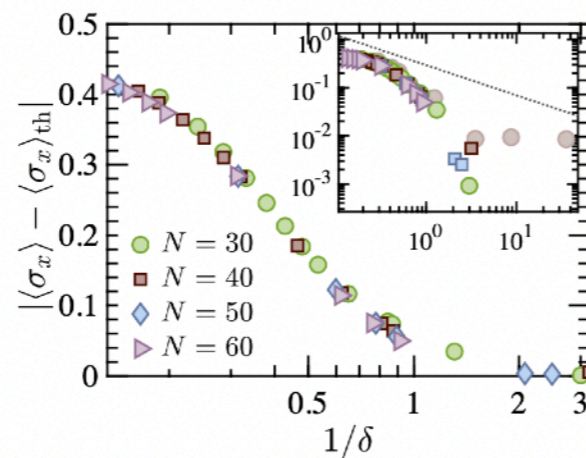
$$J = 1, g = -1.05, h = 0.5$$

### product initial states

$$|Z+\rangle := |0\rangle^{\otimes N}$$

Çakan et al., PRB103, 115113 (2021)

### local observable

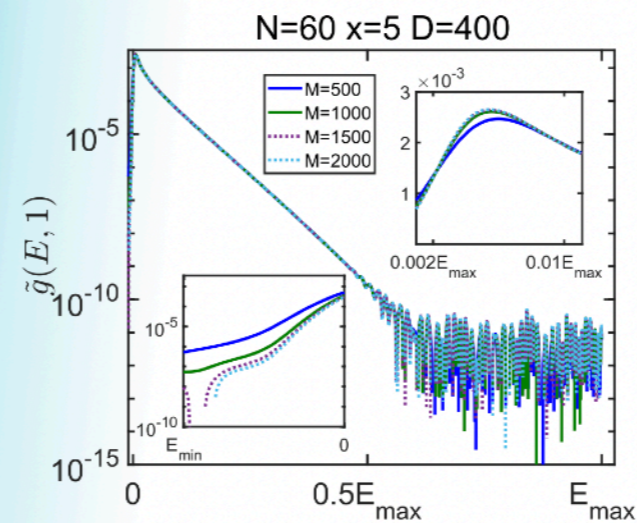


approach thermal as filter narrows

entropy in diagonal ensemble grows  $\sim$ linearly with  $N$

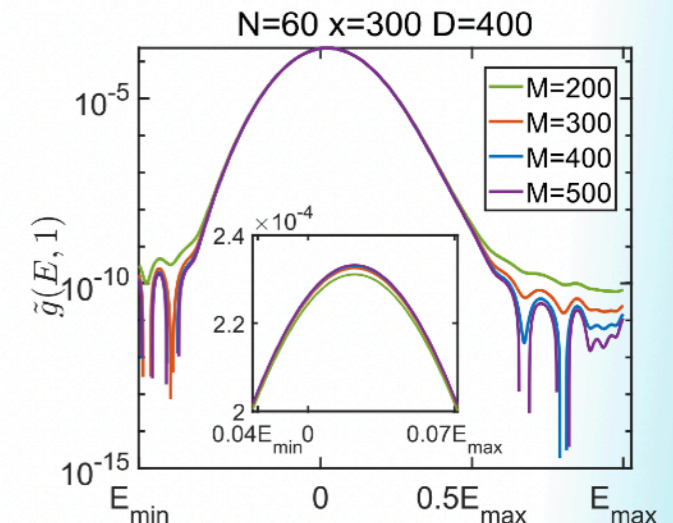
## DoS of Schwinger model

### large lattice spacing



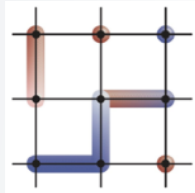
asymmetric shape

### small lattice spacing



near Gaussian shape

I. Papaefstathiou et al., PRD104, 014514 (2021)



DFG FOR 5522

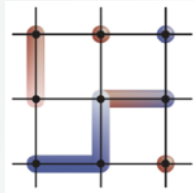


DFG TRR 360



energy filters & TNS can provide other  
(classical / quantum) tools to get dynamical  
properties





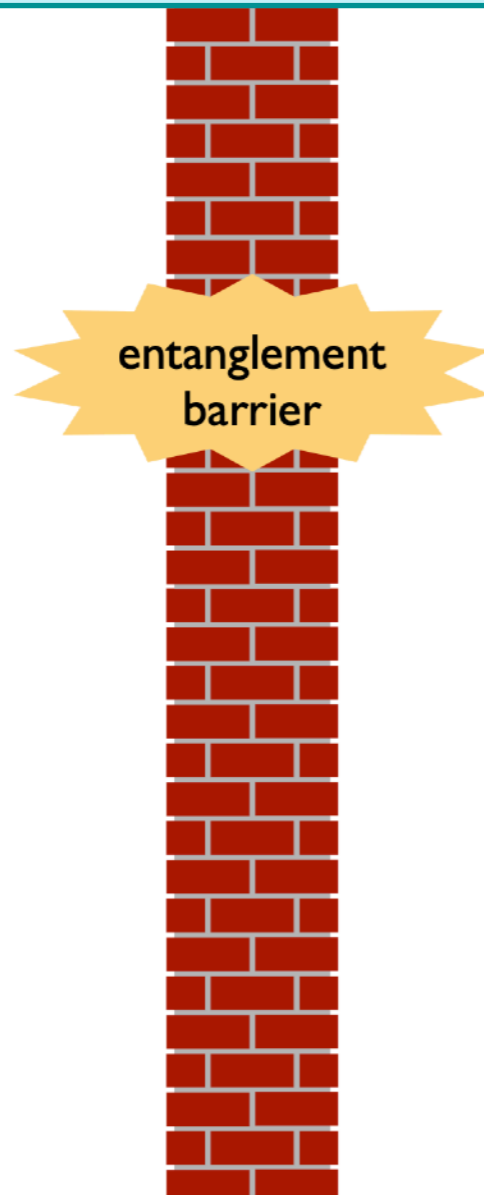
DFG FOR 5522

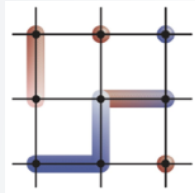


DFG TRR 360



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DFG FOR 5522



DFG TRR 360



energy filters & TNS can provide other  
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properties



entanglement  
barrier

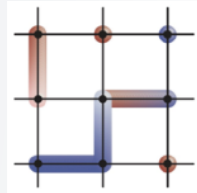
spectral properties of a  
QMB Hamiltonian

Yang, Iblisdir, Cirac, MCB, PRL 124, 100602 (2020)

Lu, PRX Quantum 2, 020321 (2021)

Yang, Cirac, MCB, PRB 106, 024307 (2022)

Luo, Trivedi, MCB, Cirac, PRB 109, 134304 (2024)



# To conclude



energy filters & TNS can provide other  
(classical / quantum) tools to get dynamical  
properties

entanglement  
barrier

spectral properties of a  
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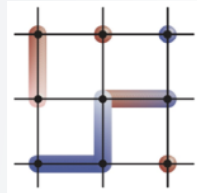
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Luo, Trivedi, MCB, Cirac, PRB 109, 134304 (2024)

further possibilities: apply to non-  
ergodic systems, probe fluctuation-  
dissipation relations, further explore  
off-diagonal matrix elements...



Thanks for your attention!



energy filters & TNS can provide other  
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entanglement  
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Yang, Iblisdir, Cirac, MCB, PRL 124, 100602 (2020)

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