

Title: Anderson localization and adiabatic quantum optimization

Date: Jul 04, 2010 11:15 AM

URL: <http://pirsa.org/10070006>

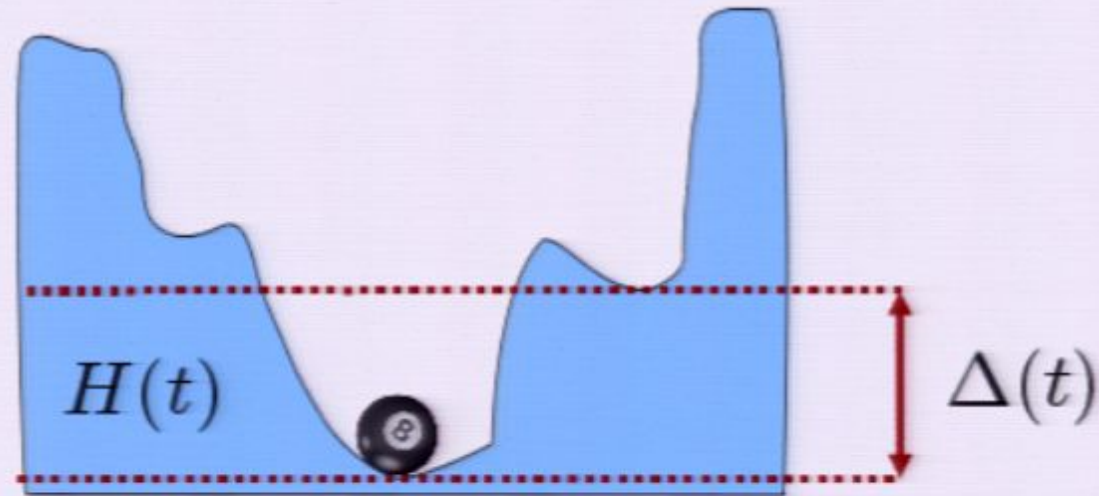
Abstract: Understanding NP-complete problems is a central topic in computer science. This is why adiabatic quantum optimization has attracted so much attention, as it provided a new approach to tackle NP-complete problems using a quantum computer. The efficiency of this approach is limited by small spectral gaps between the ground and excited states of the quantum computer's Hamiltonian.

We show that the statistics of the gaps can be analyzed in a novel way, borrowed from the study of quantum disordered systems in statistical mechanics. It turns out that due to a phenomenon similar to Anderson localization, exponentially small gaps appear close to the end of the adiabatic algorithm for large random instances of NP-complete problems. We show that this effect makes adiabatic quantum optimization fail, as the system gets trapped in one of the numerous local minima. We will also discuss recent developments including the effect of the exponential number of solutions and Hamiltonian path change.

Joint work with Boris Altshuler and Hari Krovi

Based on arXiv:0908.2782 and arXiv:0912.0746

Adiabatic evolution



Slowly varying $H(t) \rightarrow$ Stays close to ground state

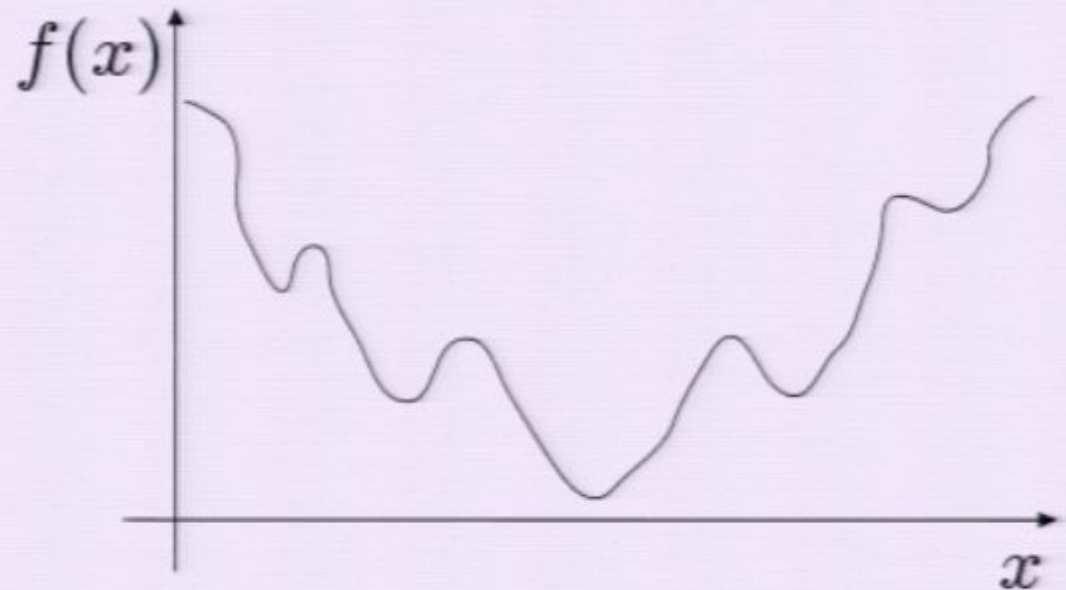
Probability of excitation depends on

- Total time T (slower is better)
- Gap $\Delta(t)$ (larger gap is better)

Adiabatic quantum optimization

[Farhi *et al.* '00]

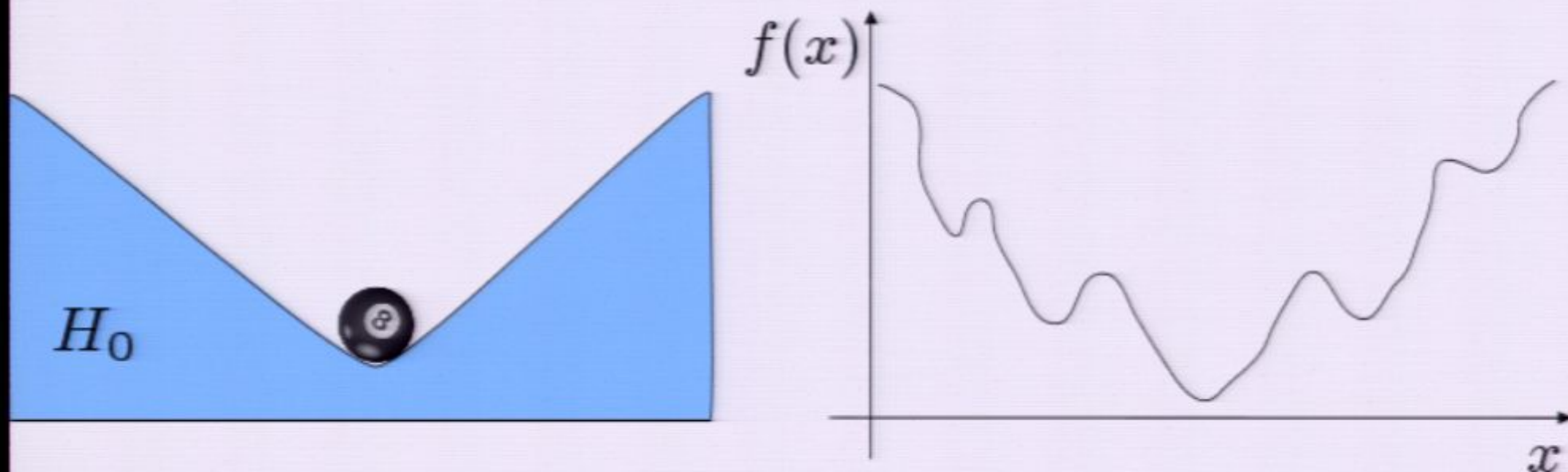
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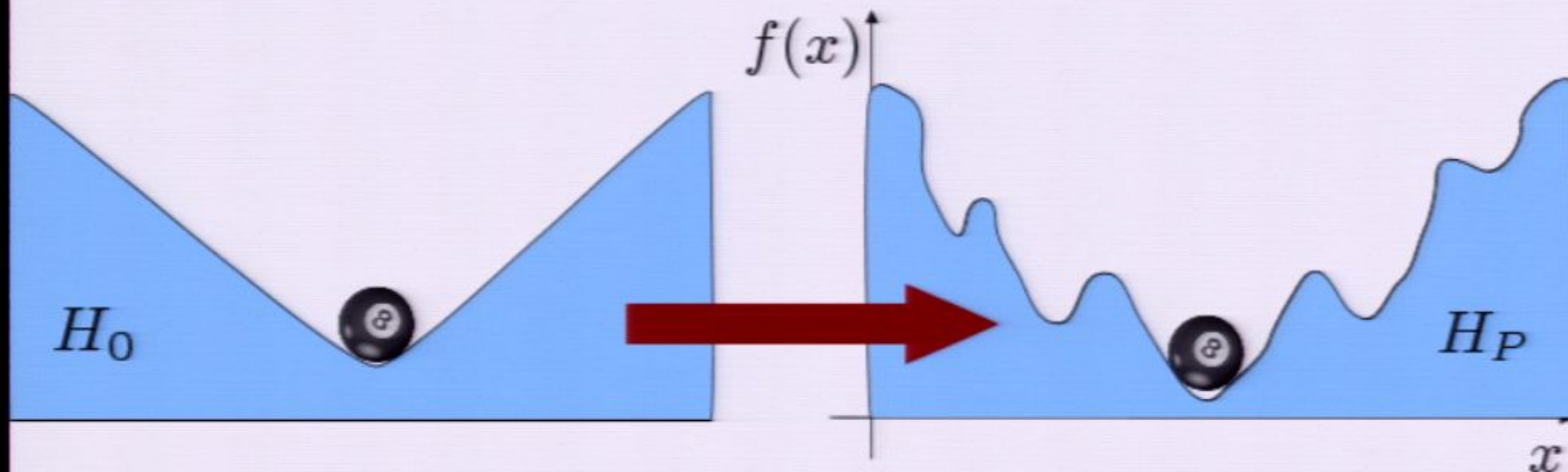
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Adiabatic quantum optimization

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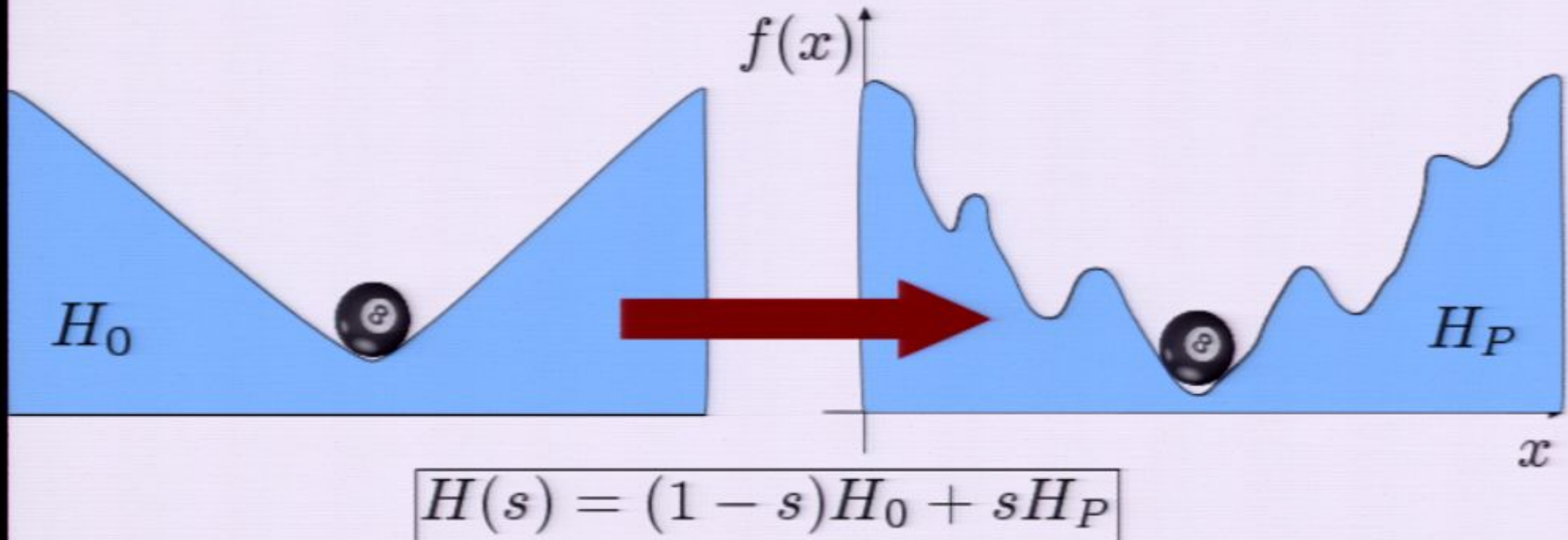
$$H(s) = (1 - s)H_0 + sH_P$$

where $s(t) = t/T$

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How powerful is it?

- It is quantum! Unstructured search in time $O(\sqrt{N})$ (cf Grover)

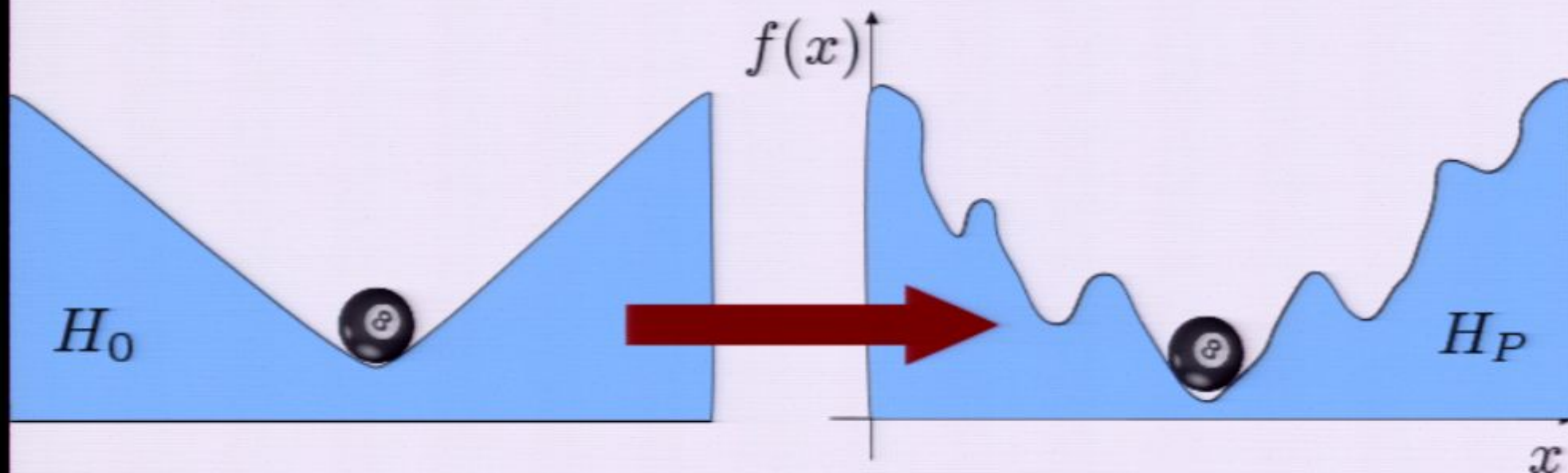


[vanDam-Mosca-Vazirani'01, Roland-Cerf'02]

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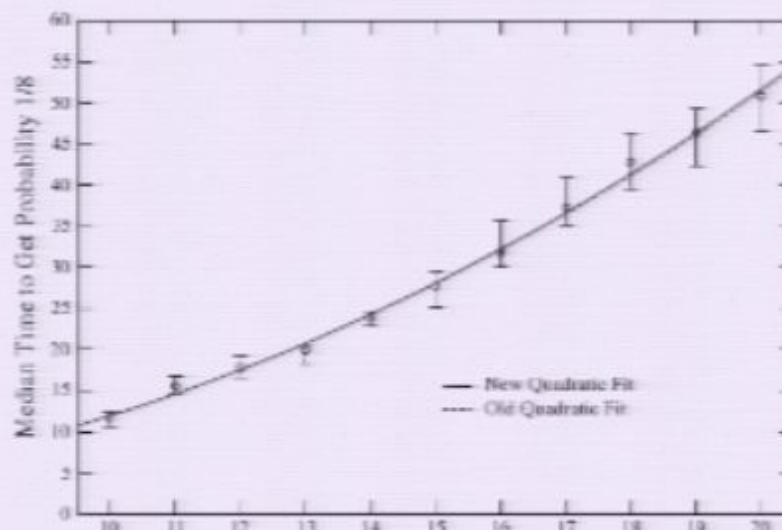
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[Farhi *et al.*'00, Hogg'03, Banyuls *et al.*'04, Young *et al.*'08]



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- for bad choice of initial Hamiltonian

[Znidaric-Horvat'06, Farhi *et al.*'08]

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But maybe typical gaps are only polynomial?

Exact-Cover 3 (EC3)

- NP complete problem (similar to 3-SAT)

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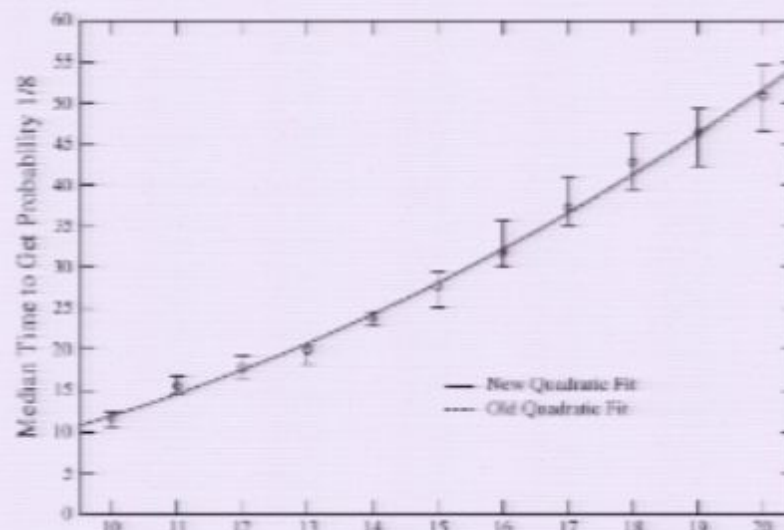
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- M clauses of 3 bits:

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$$= M - \sum_i B_i x_i + 2 \sum_{i,j} J_{ij} x_i x_j$$

#clauses

#clauses with bit i

#clauses with bits i, j

Random instances

- Pick M clauses uniformly at random

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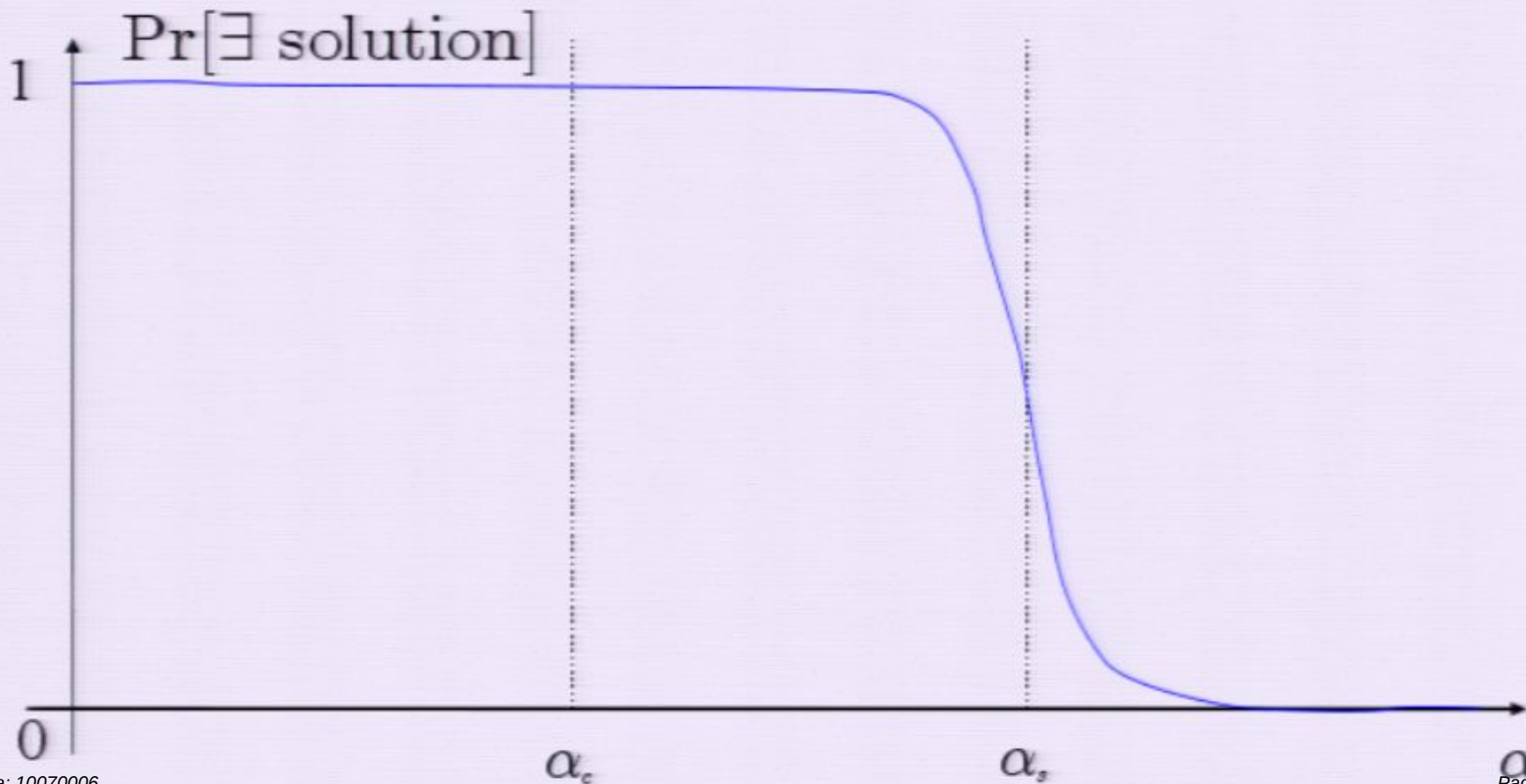
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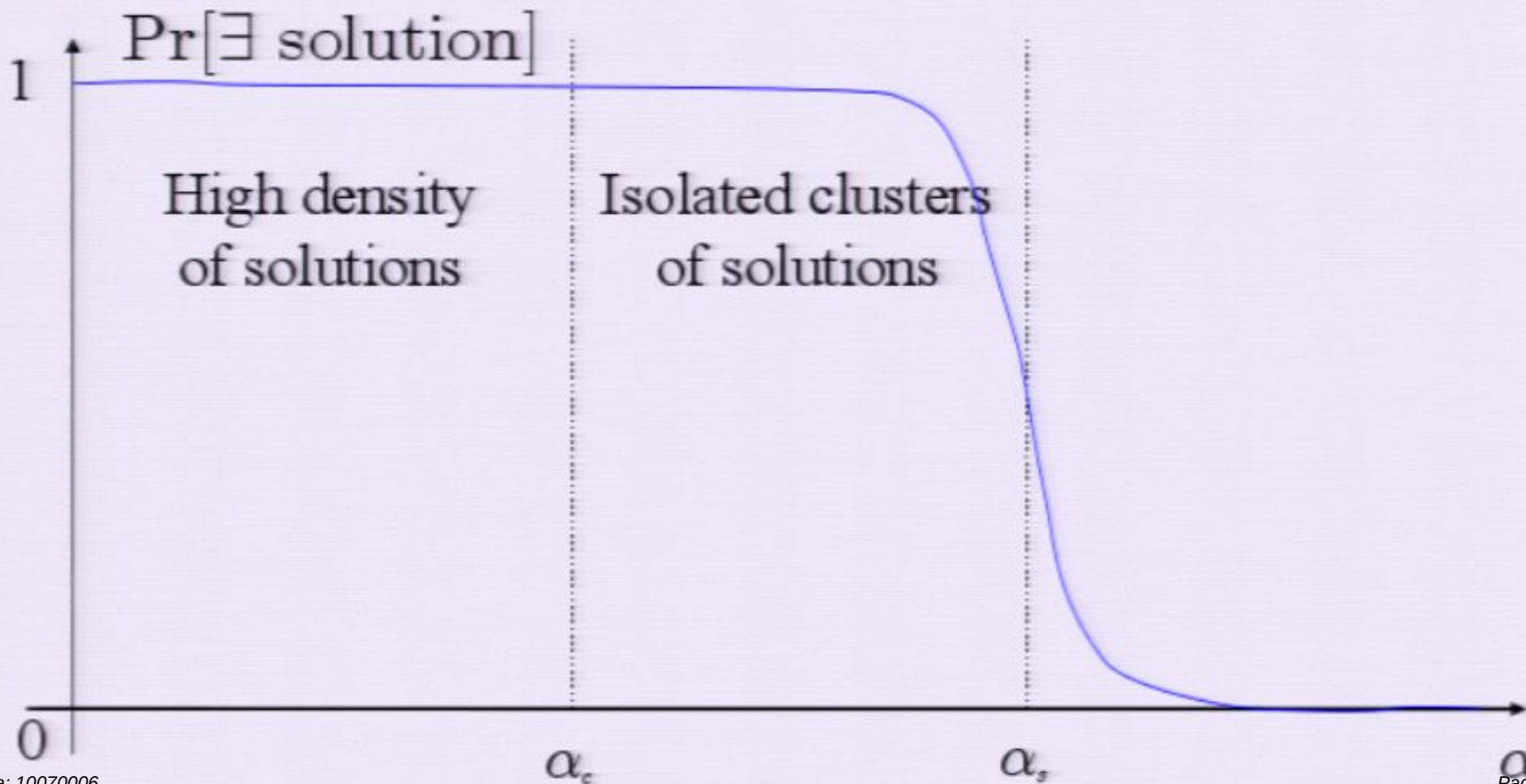
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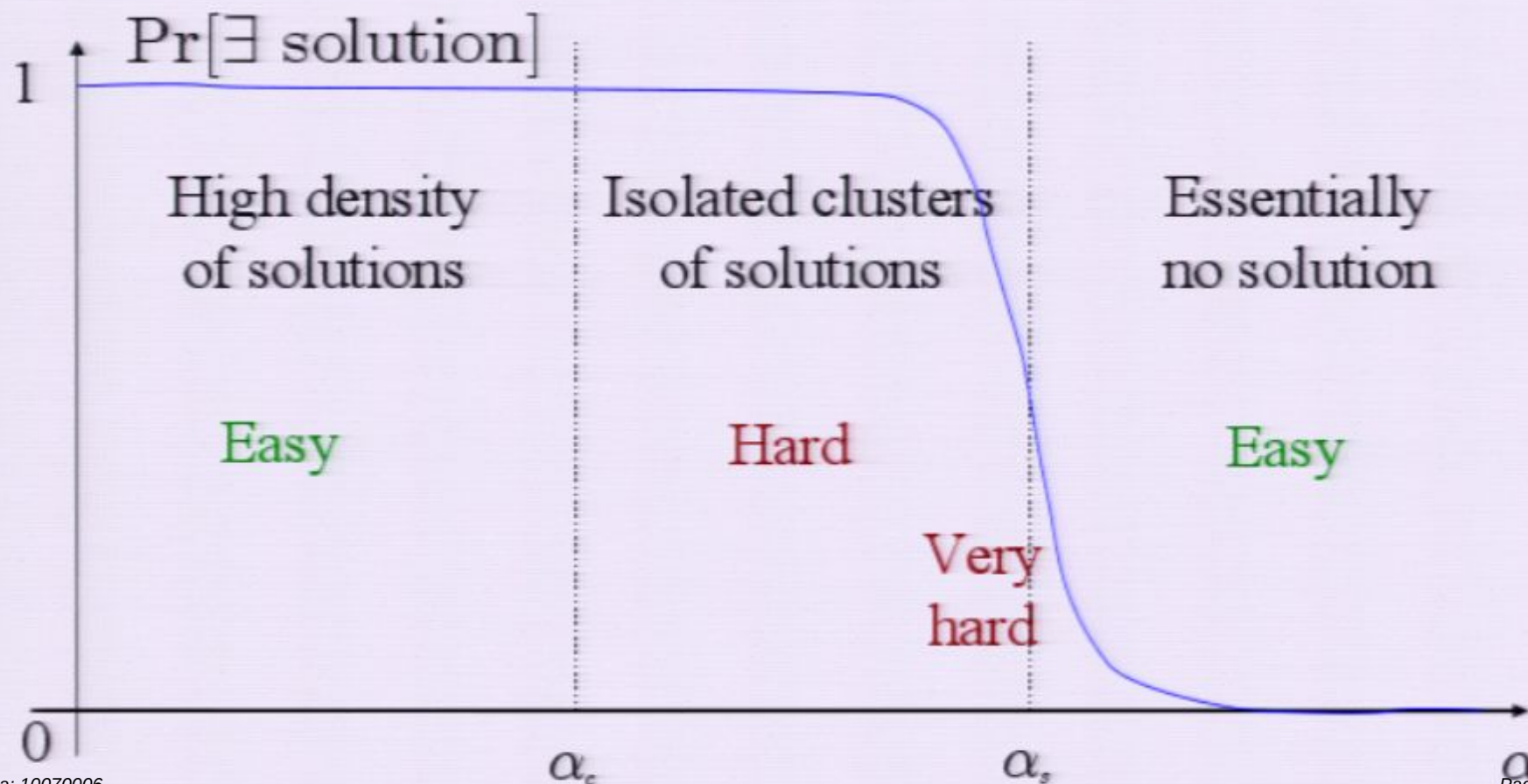
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Connection to Anderson localization

To study the spectrum of $H(s)$ close to $s=1$, we consider

$$H(\lambda) = \frac{H(s)}{s} = H_P + \lambda H_0 \quad \text{where} \quad \lambda = \frac{1-s}{s}$$

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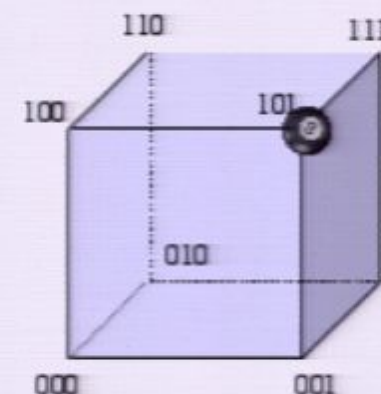
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$\Rightarrow$ Particle hopping on a hypercube



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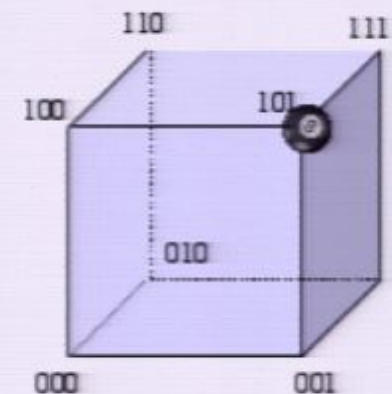
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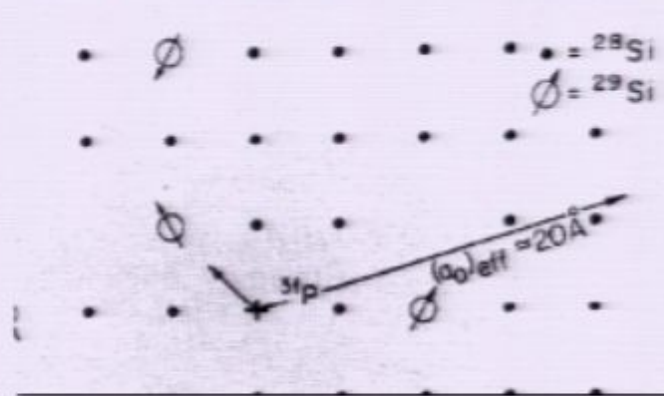
$\Rightarrow$ Particle hopping on a hypercube

$\Rightarrow$ Similar to Anderson's tight binding model



Anderson localization

“Extended states become localized due to disorder”



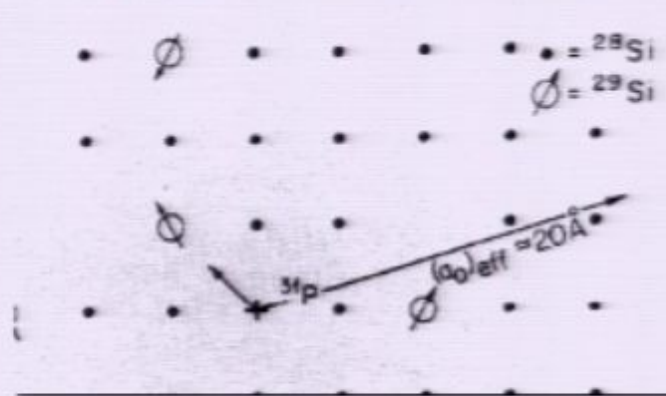
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Nobel Prize
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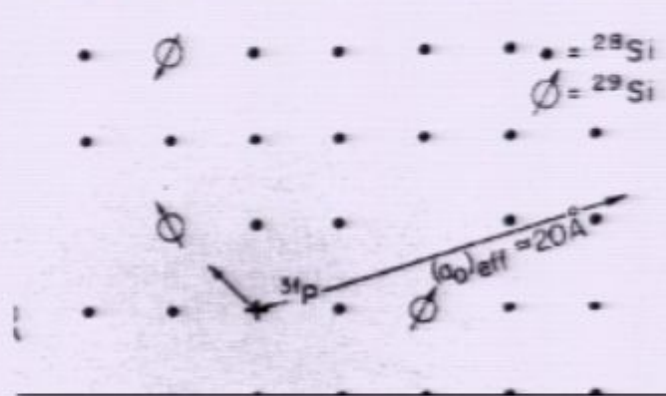
- Grid with coupling λ
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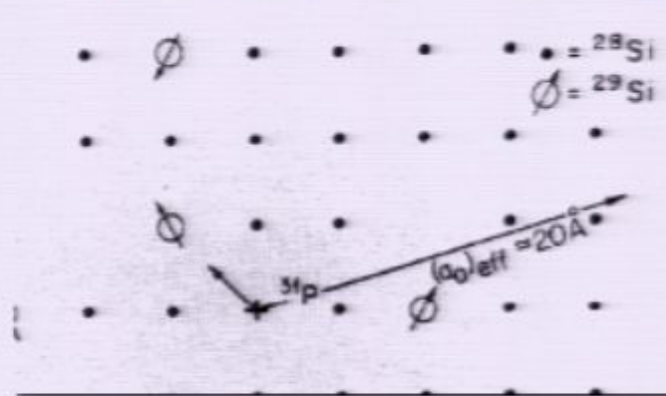
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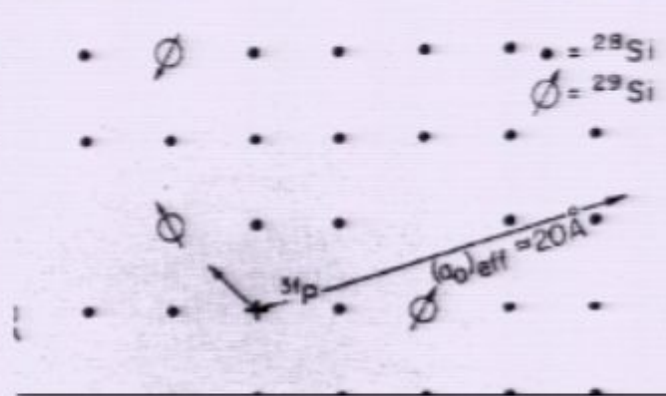
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$\lambda > \lambda_c \rightarrow$ Extended state $\rightarrow$ Metal

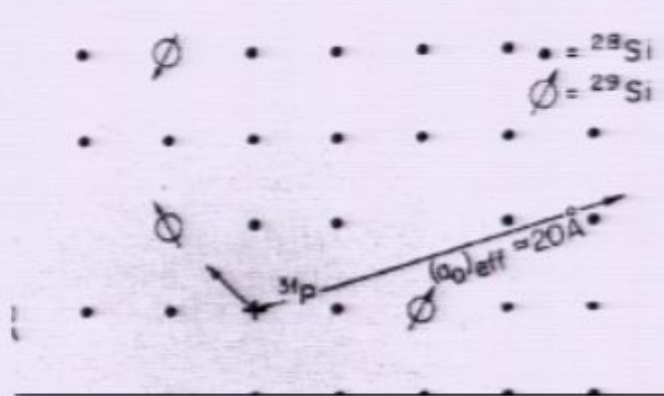
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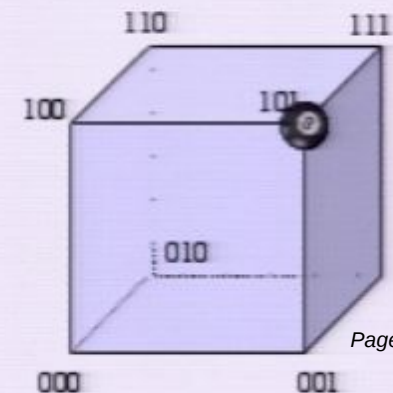
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In our case:

- Hypercube with coupling λ
- Energies from random Exact-Cover 3

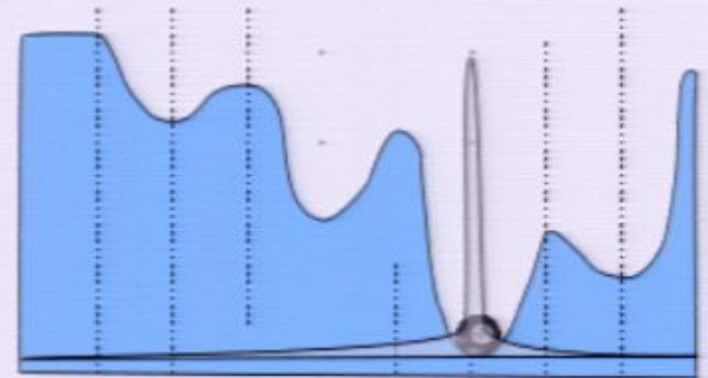


Localized and extended states

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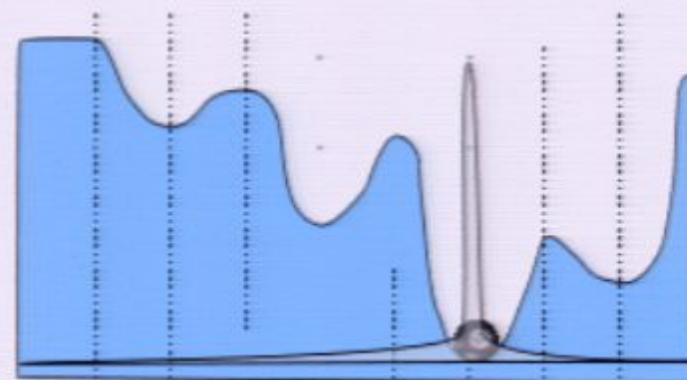
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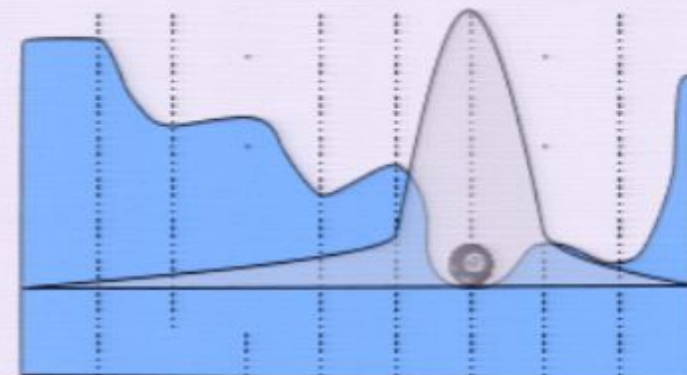
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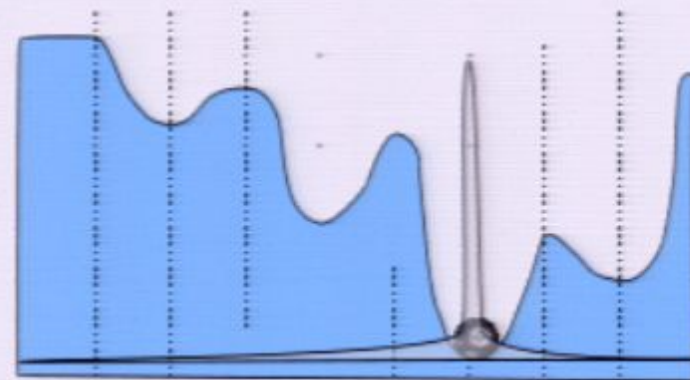
- Transverse field “spreads” the state



Localized and extended states

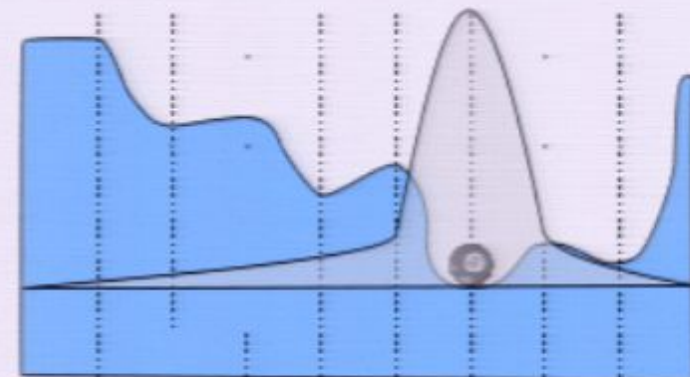
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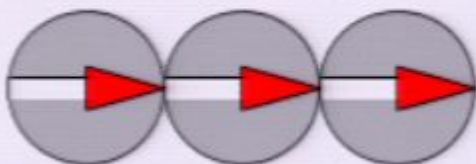
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$$\lambda \gg 1$$

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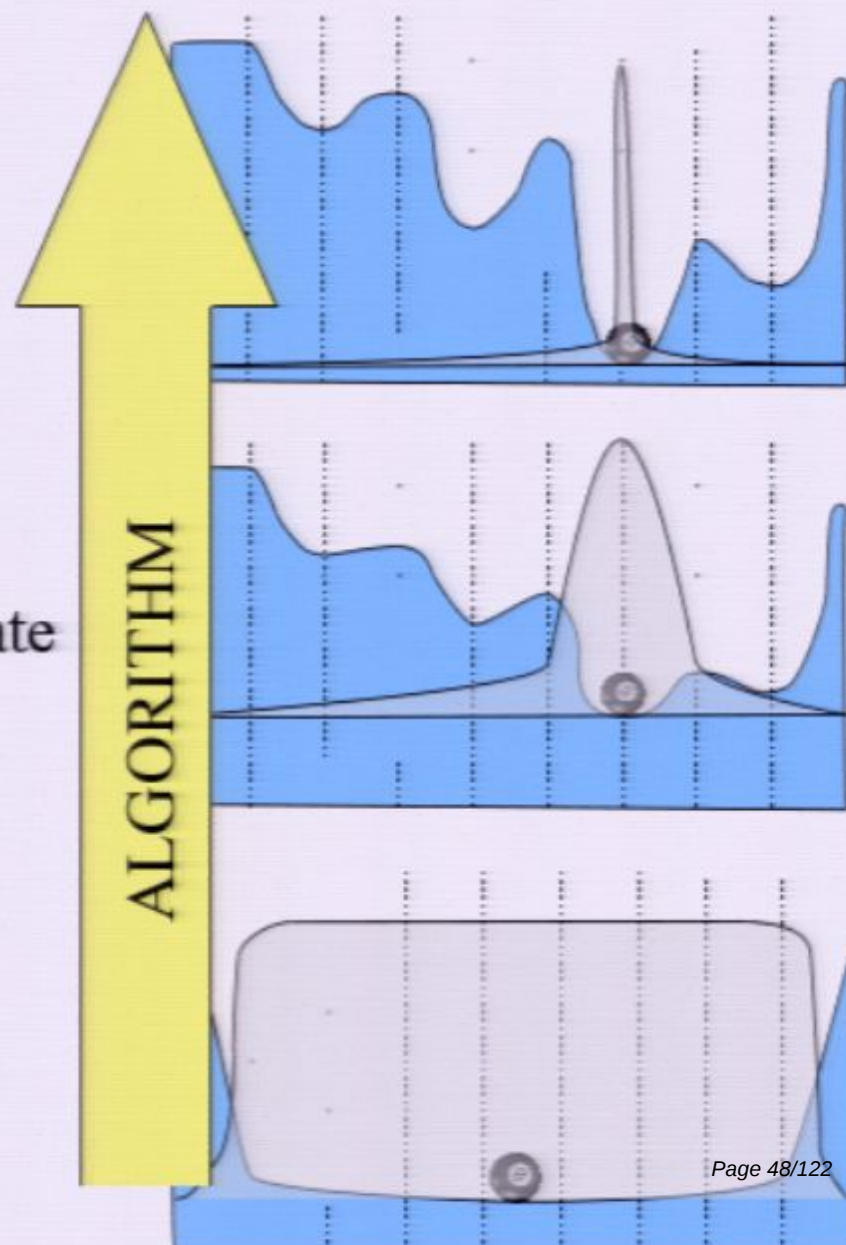
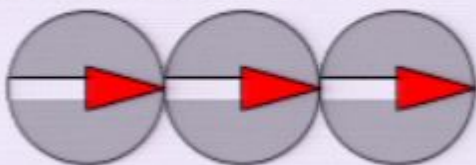
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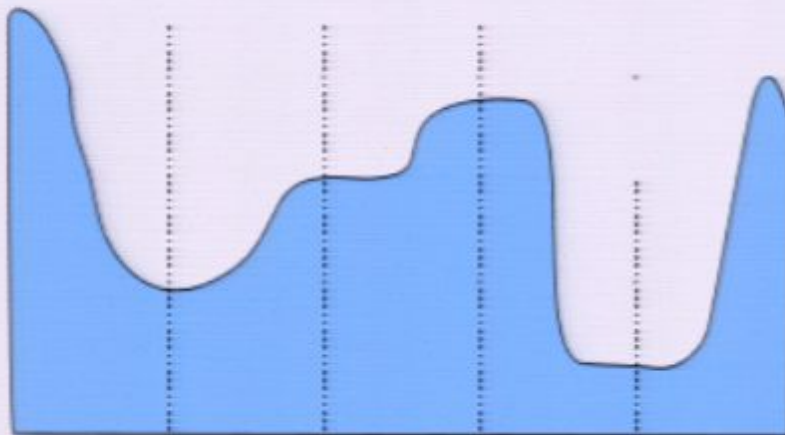
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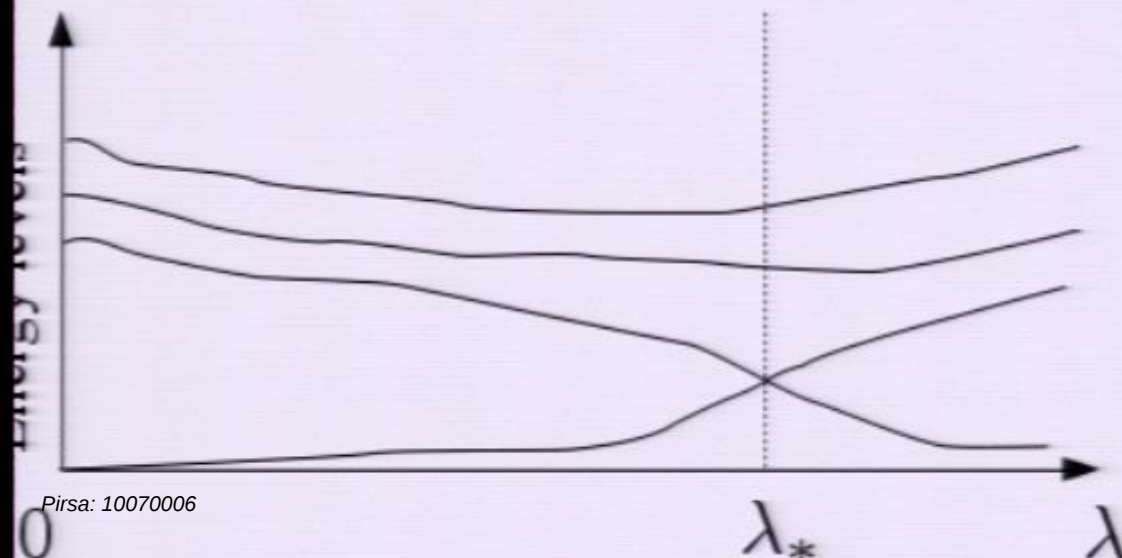
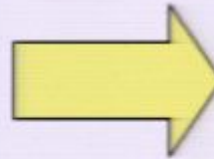
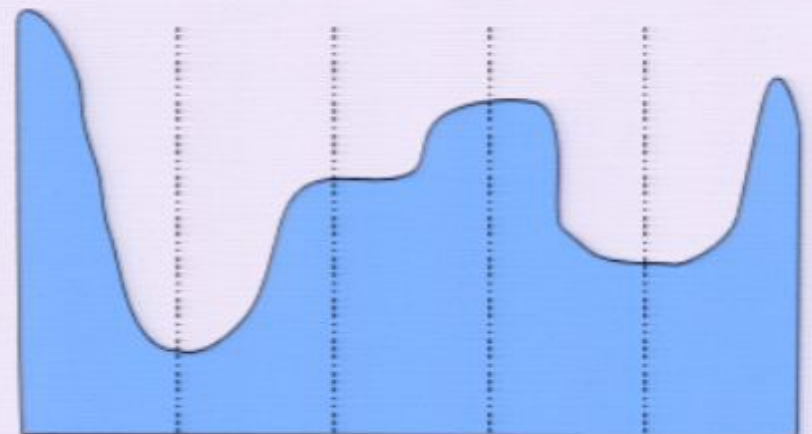
Tunneling: extended state

What if a local minimum later becomes the global minimum?

$$\lambda > \lambda_*$$



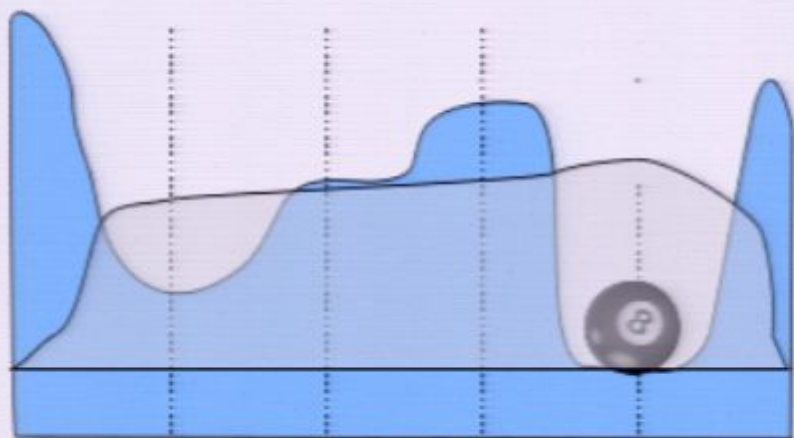
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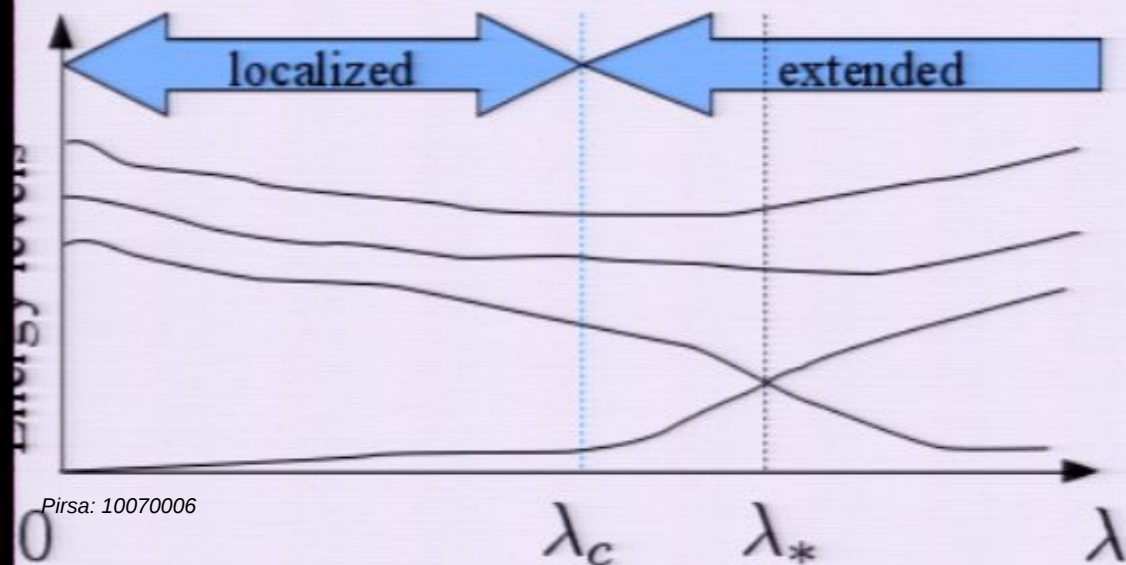
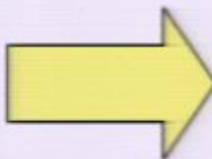
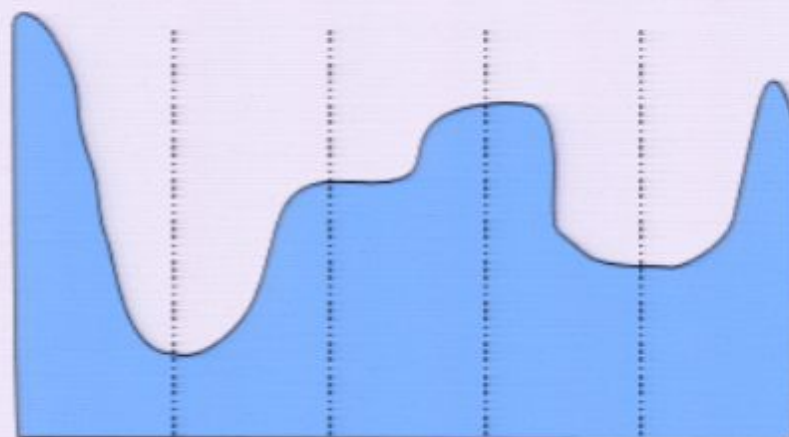
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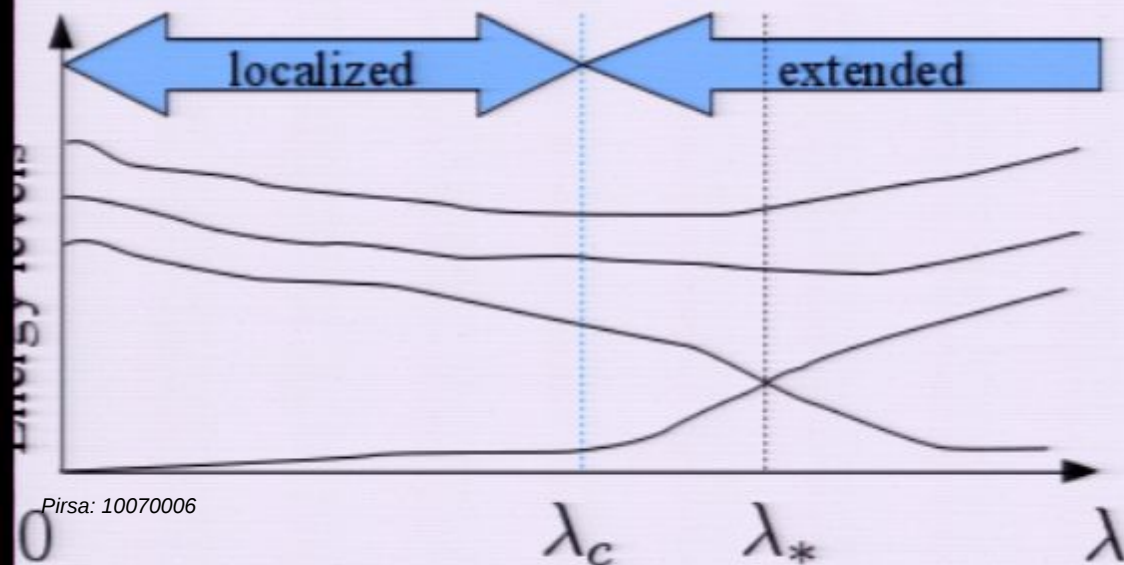
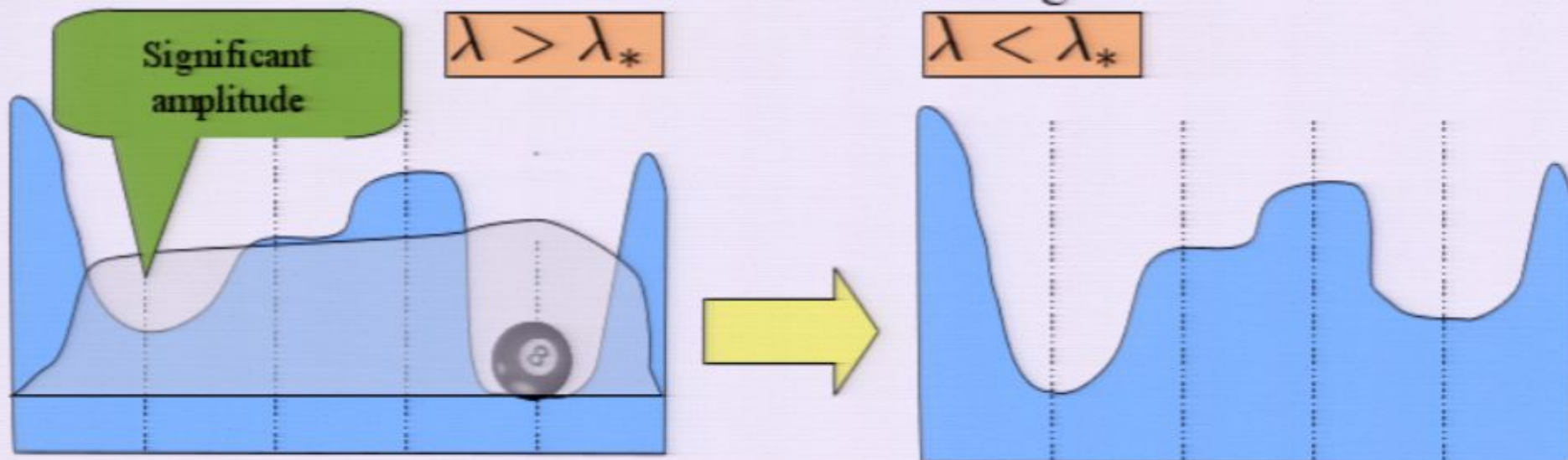


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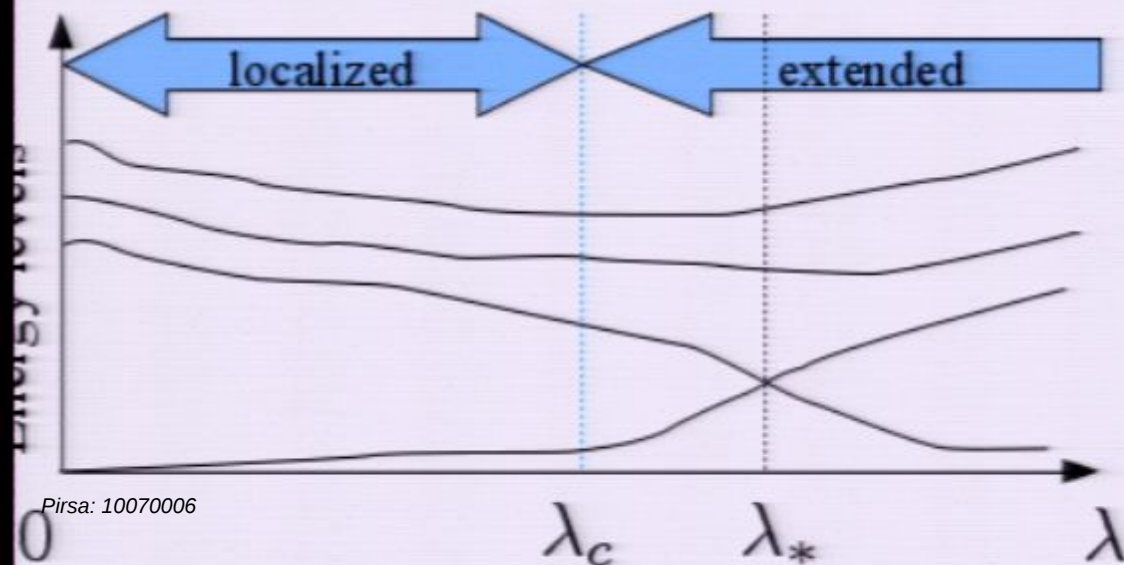
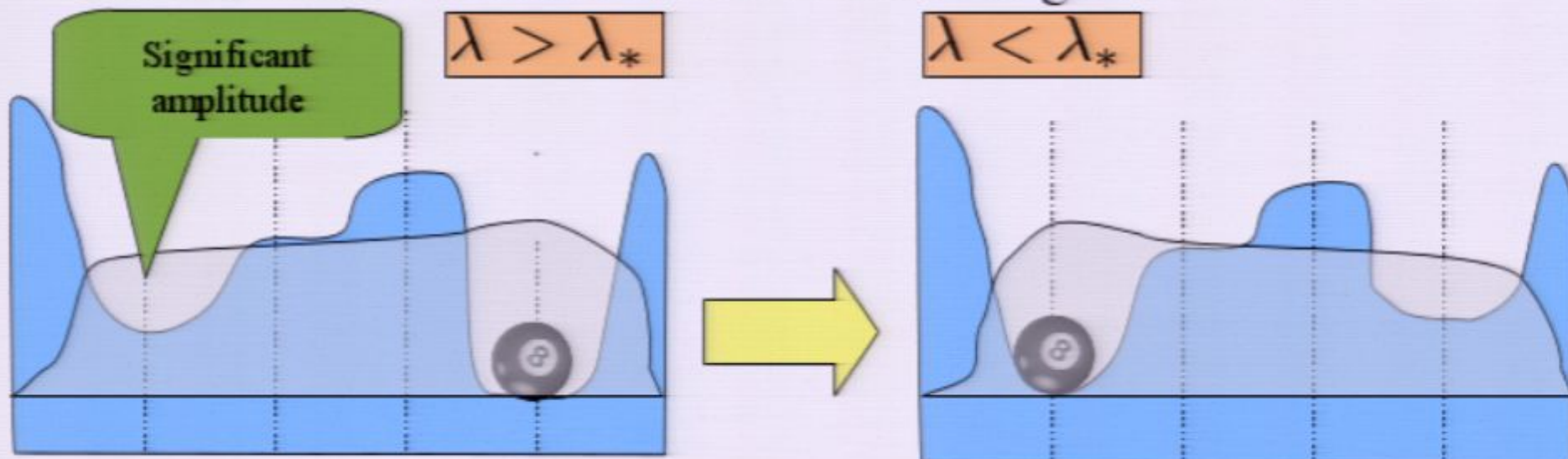
Tunneling: extended state

What if a local minimum later becomes the global minimum?



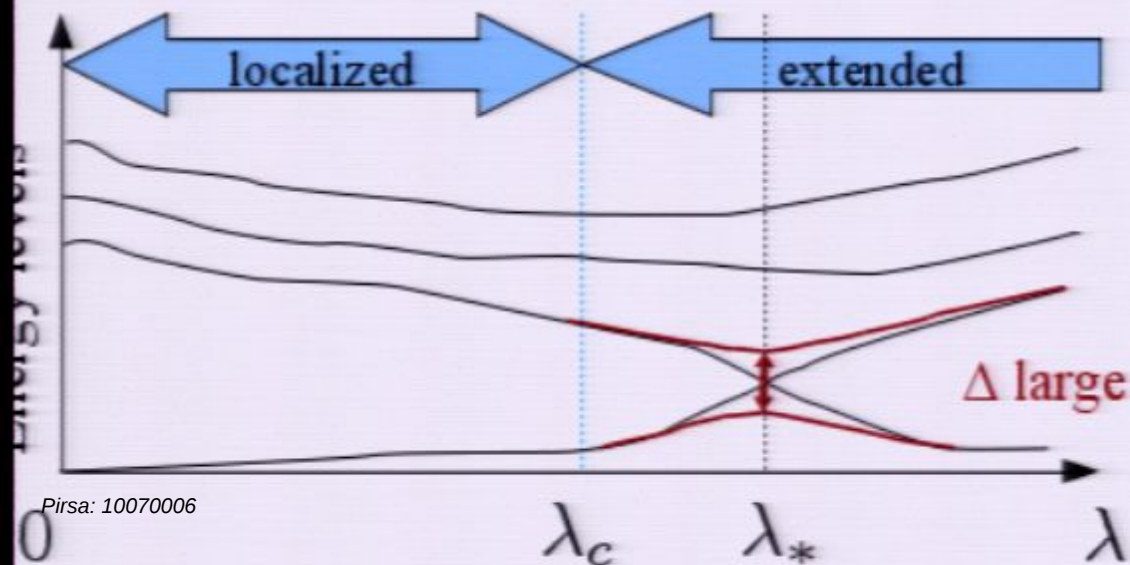
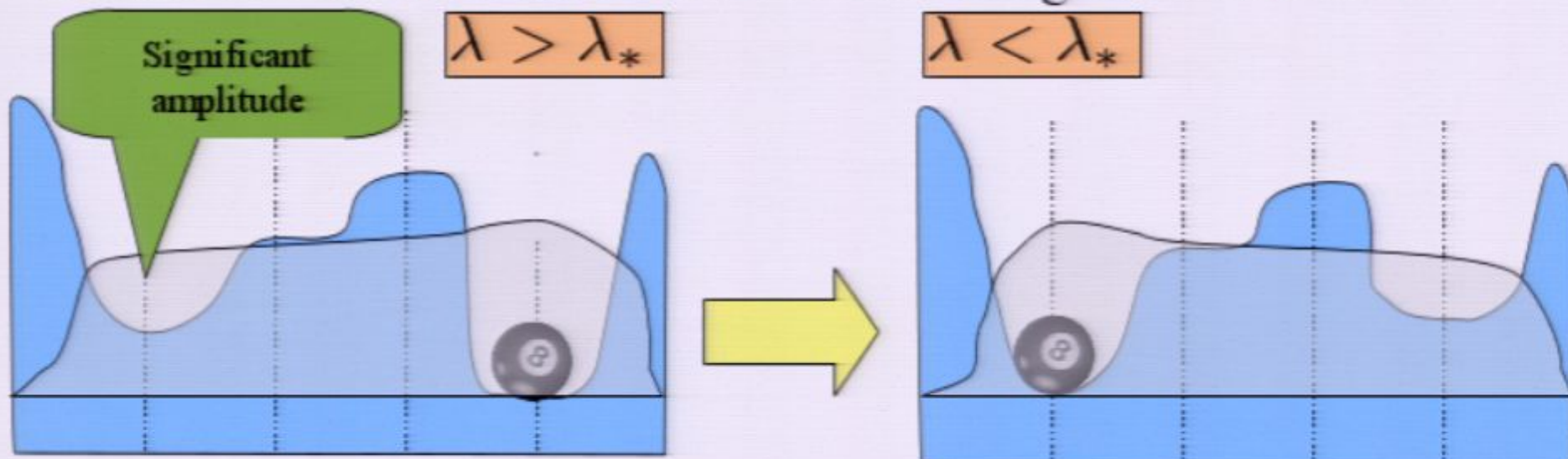
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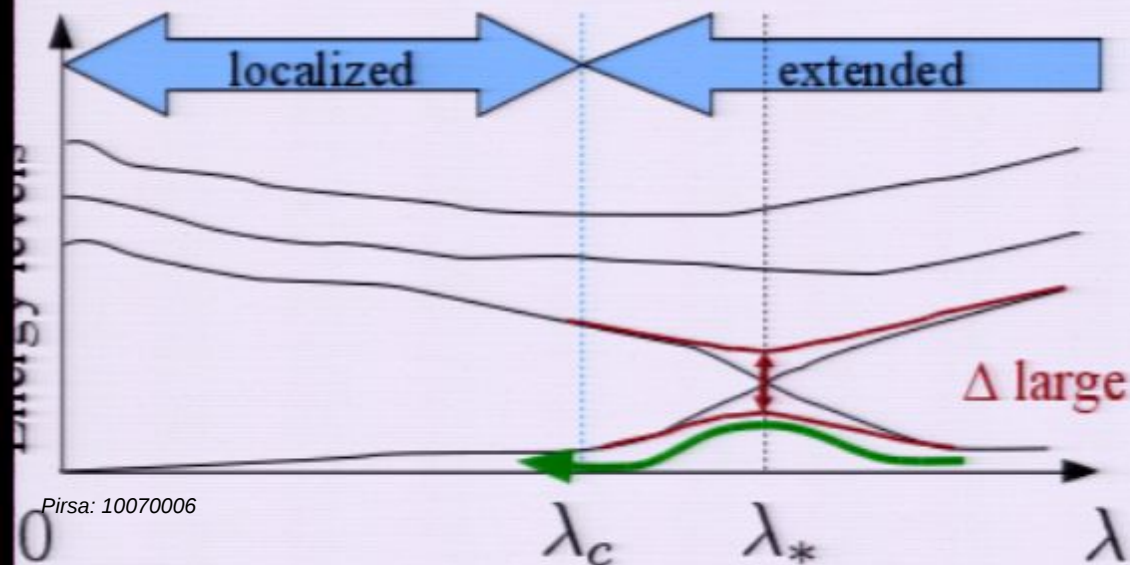
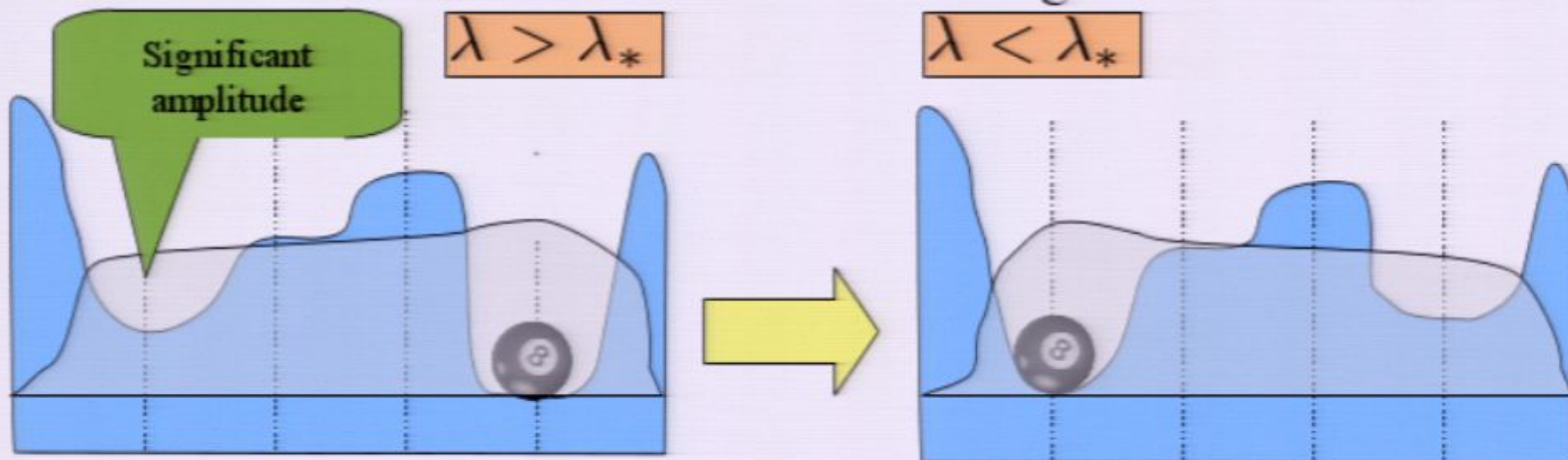
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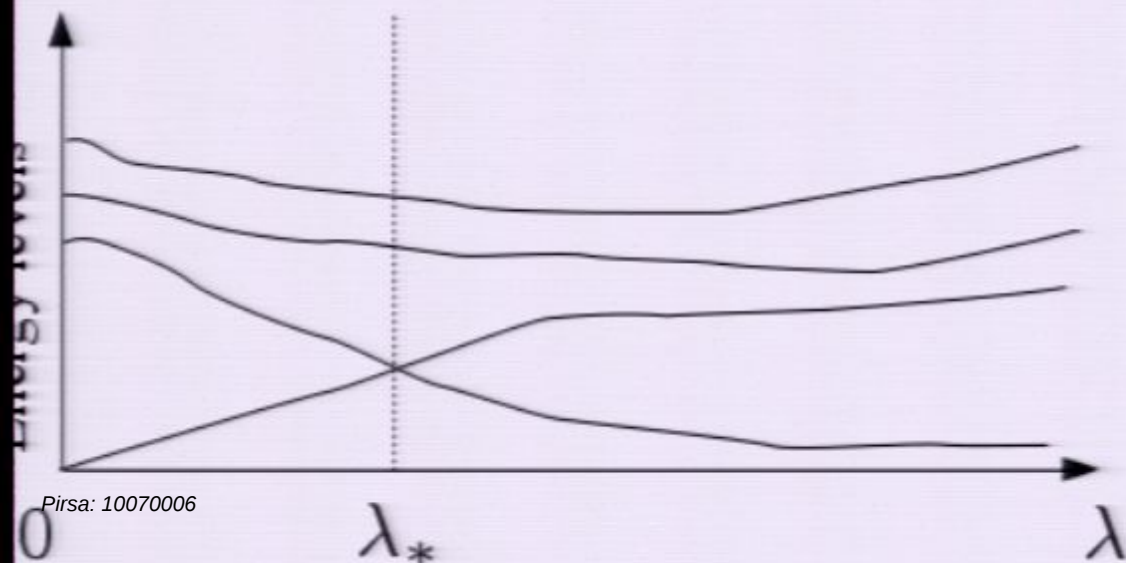
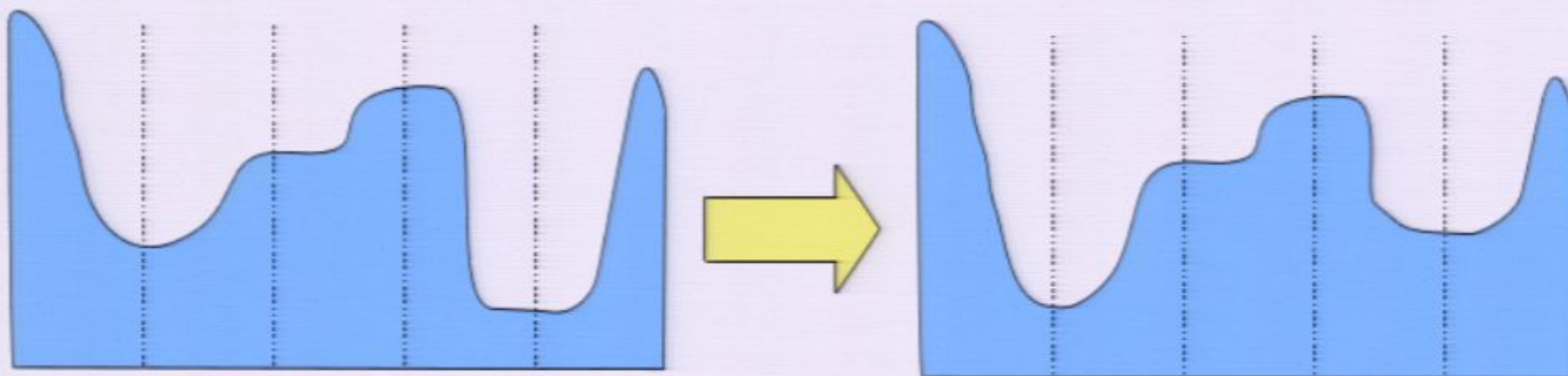


Tunneling: localized state

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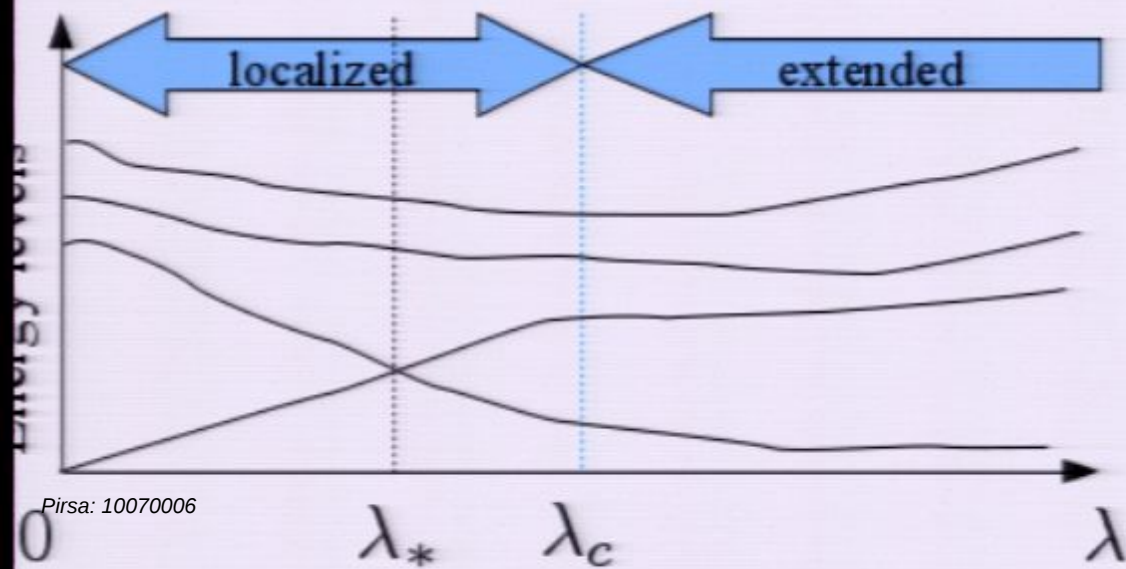
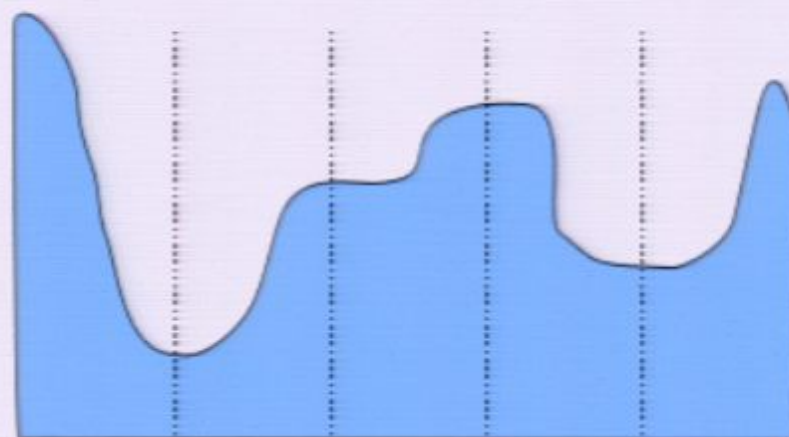
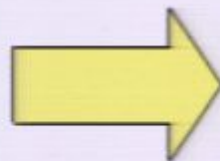
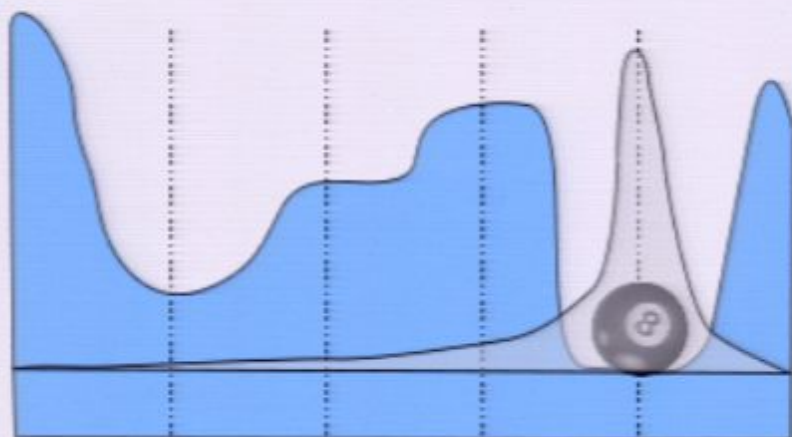


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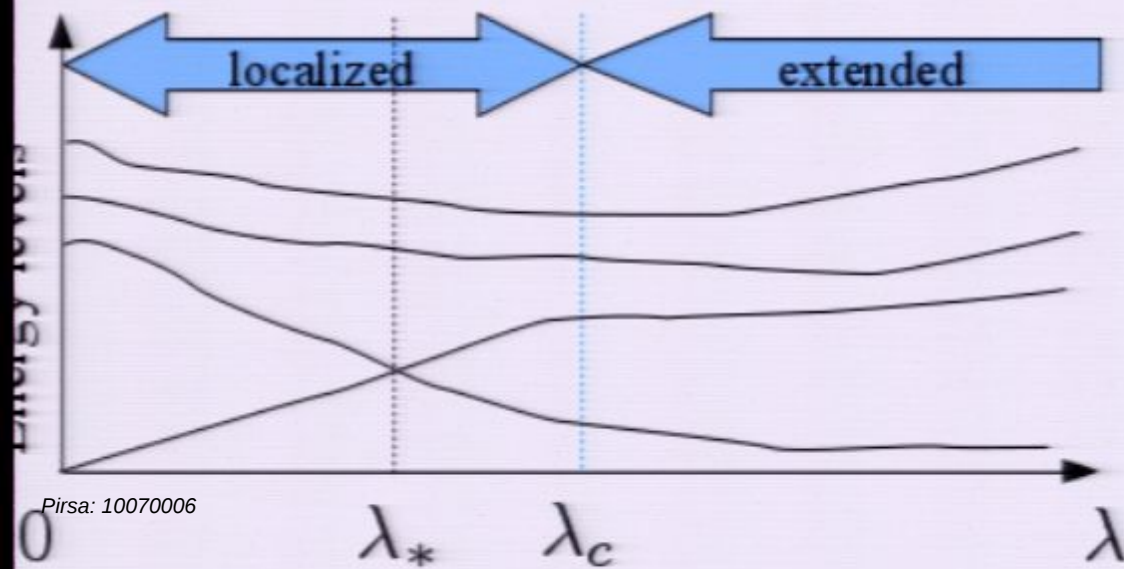
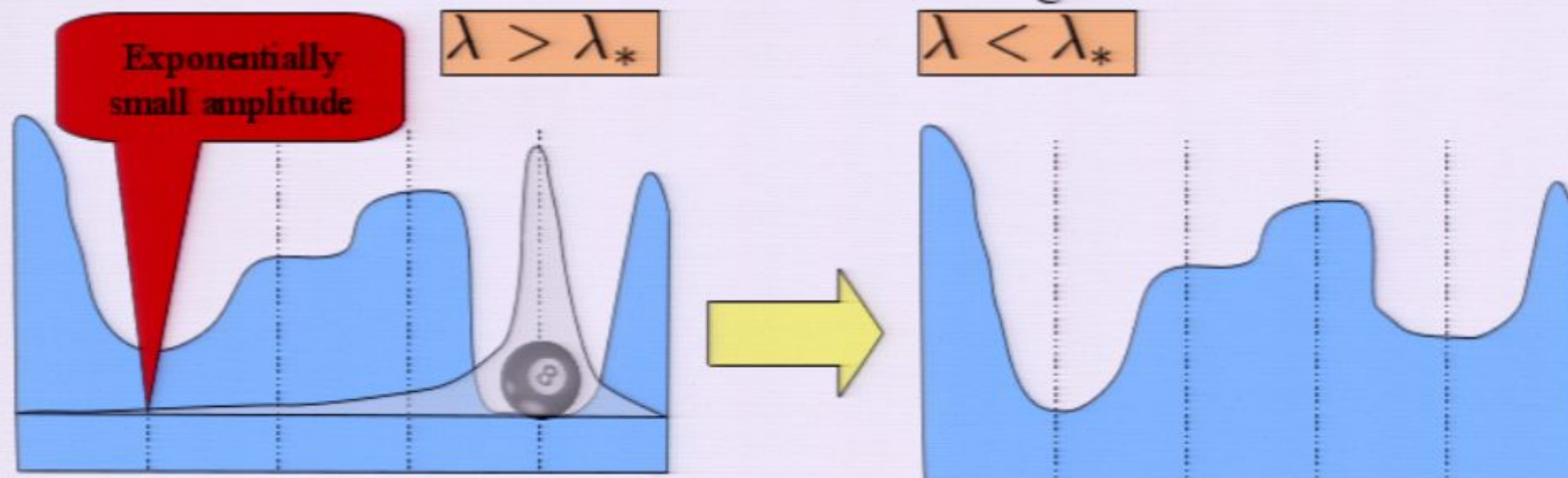
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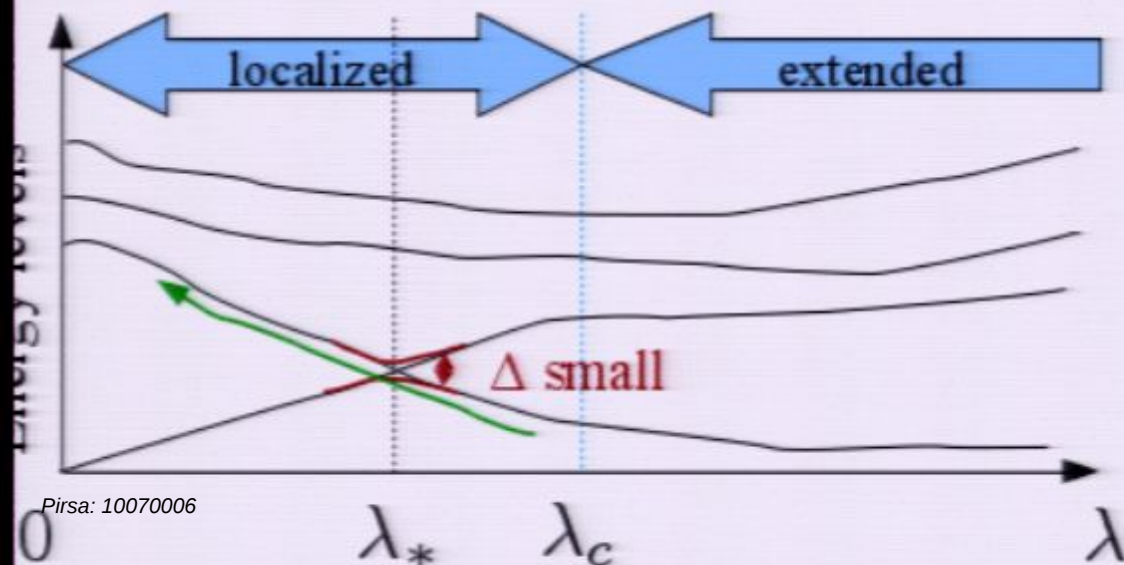
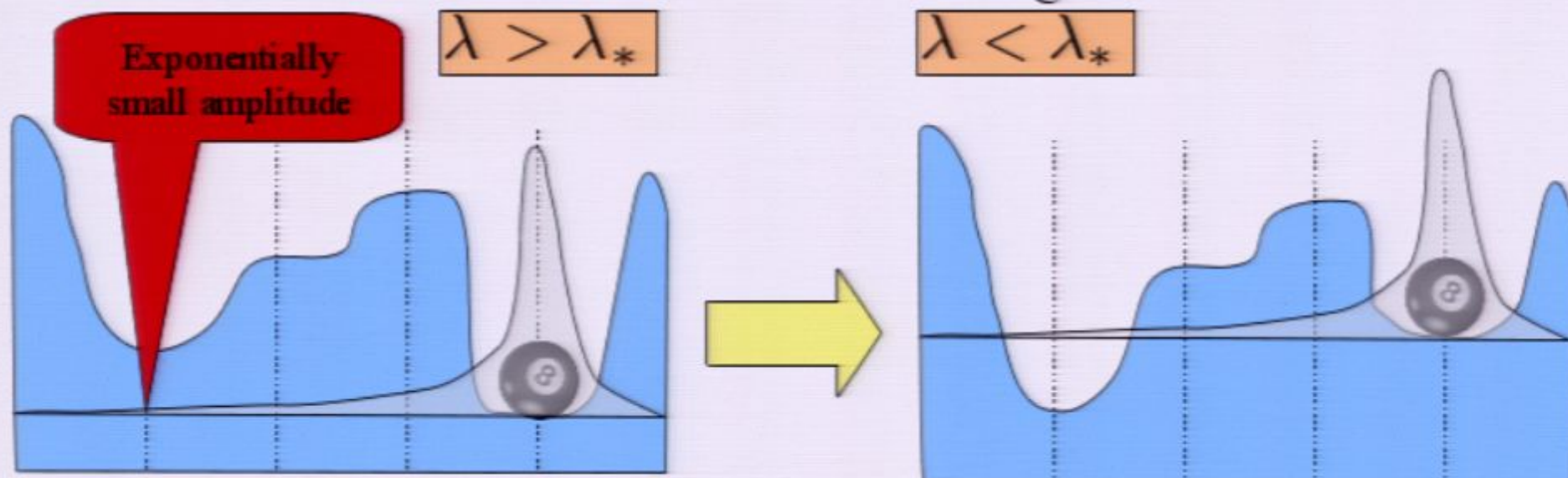
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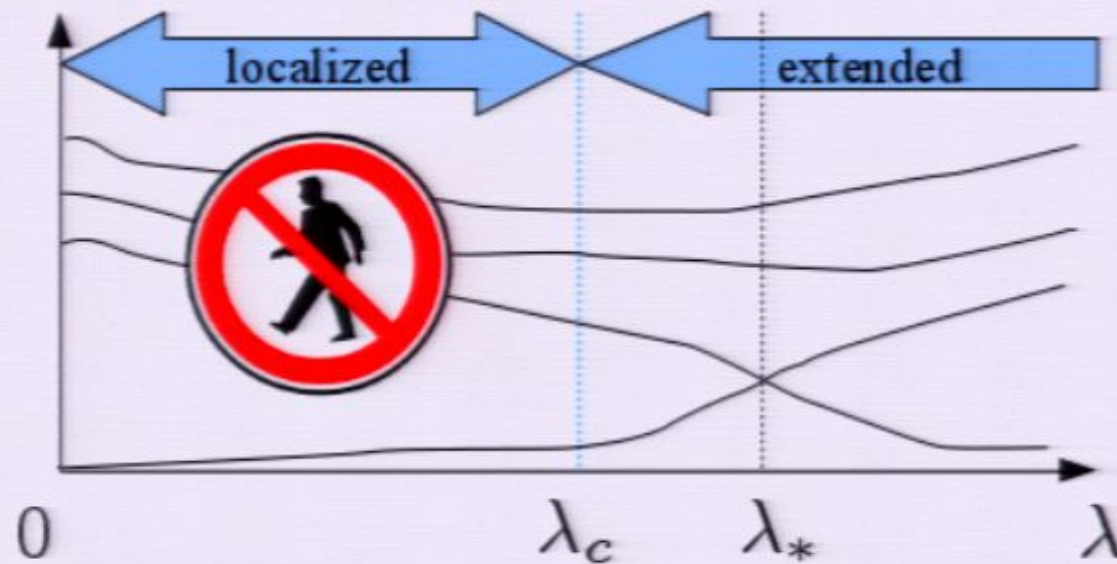
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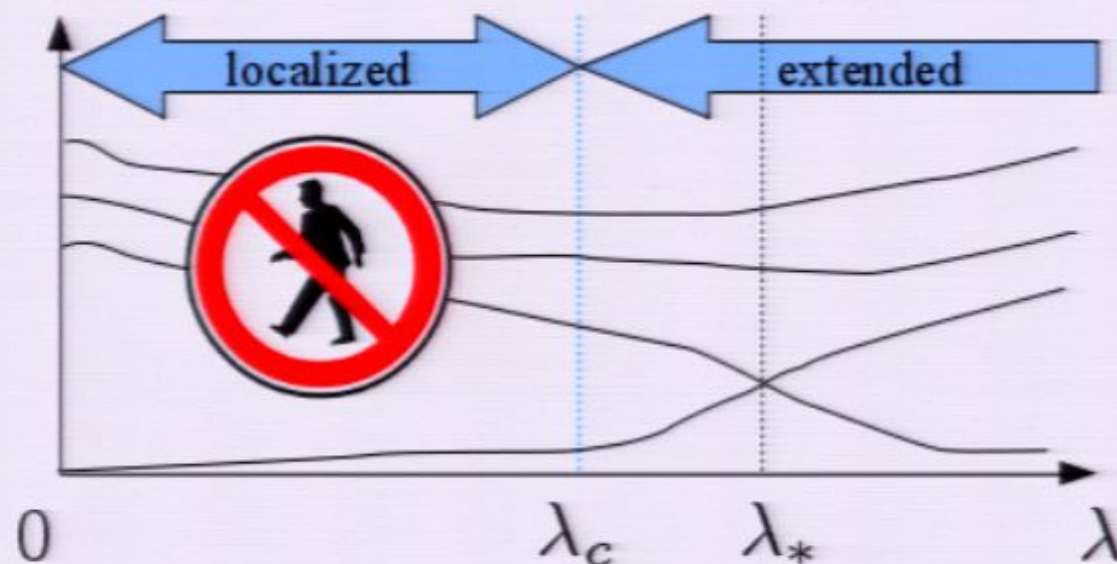


Our result



As the size of the problem N increases

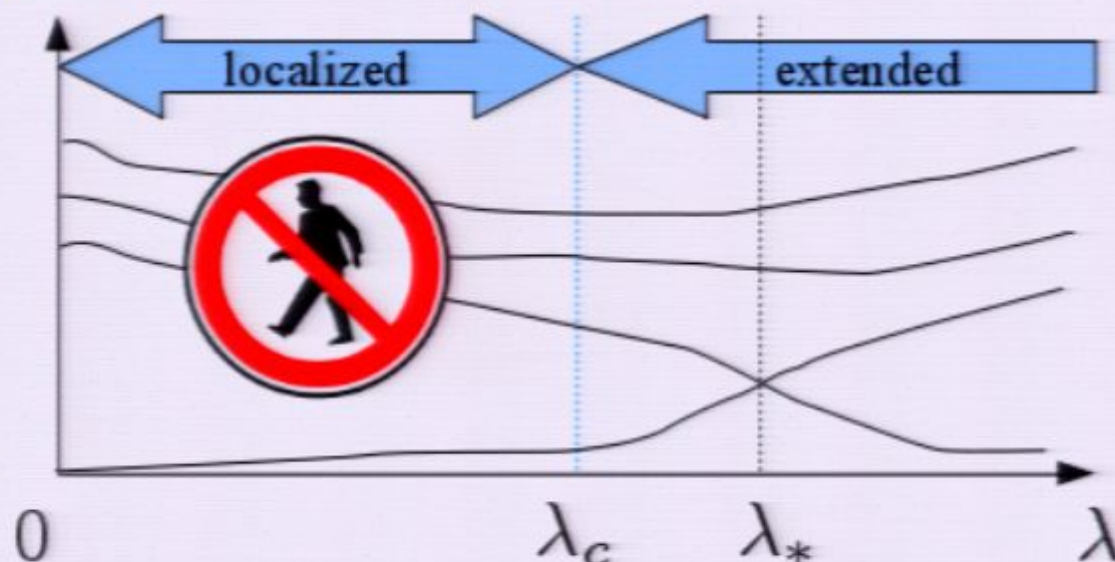
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1) Anderson localization would imply $\lambda_c = \Omega(1/\log N)$

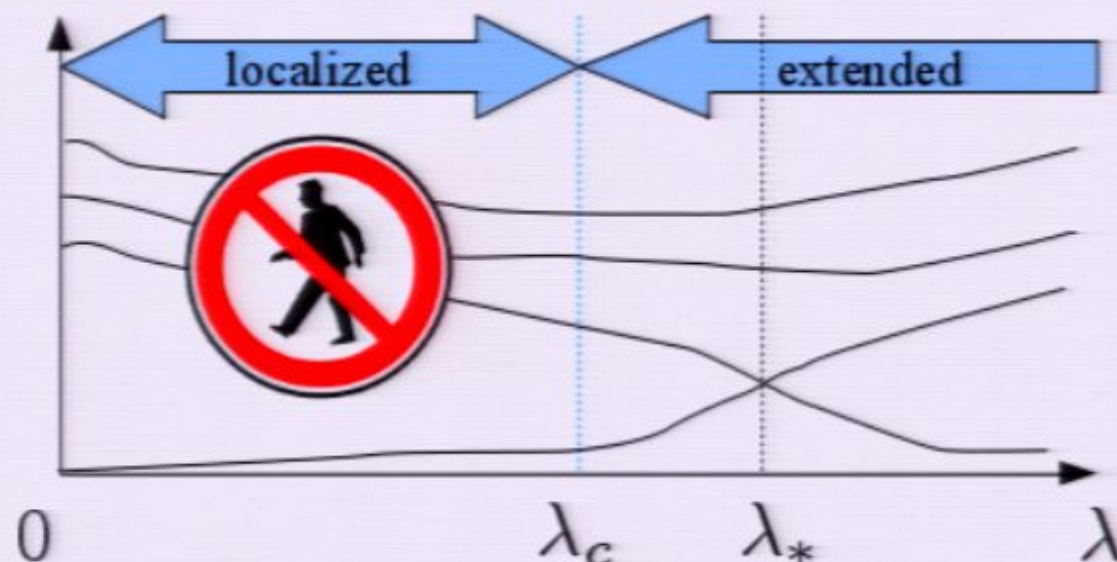
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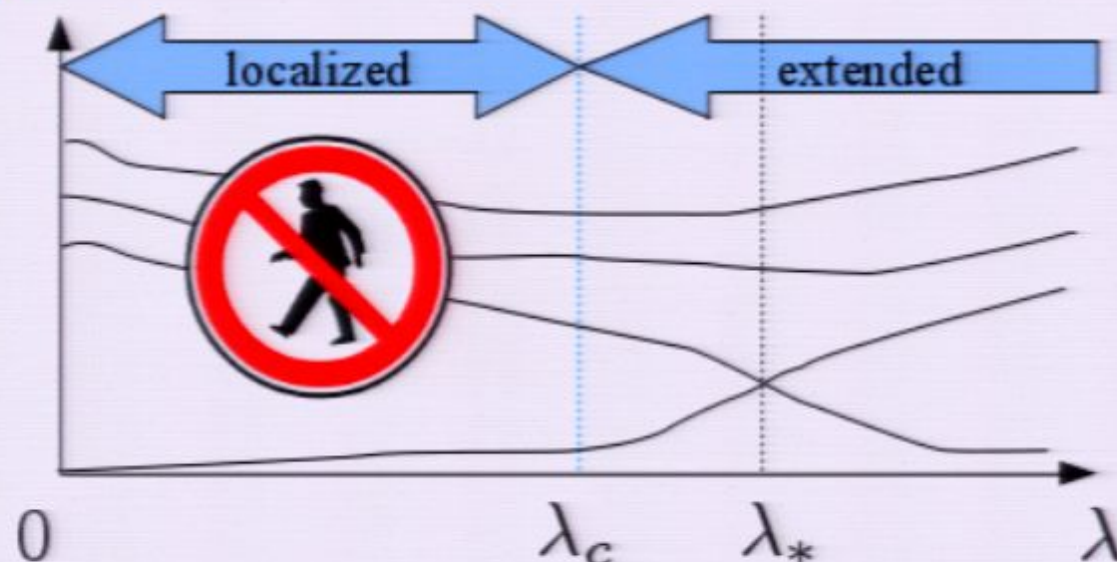


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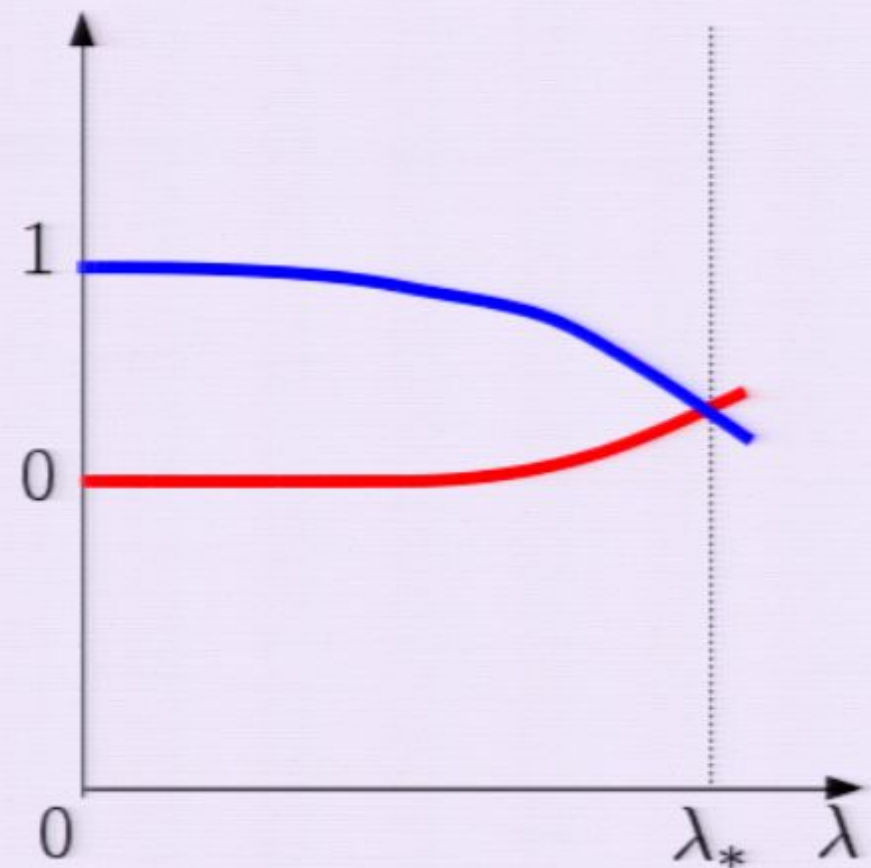
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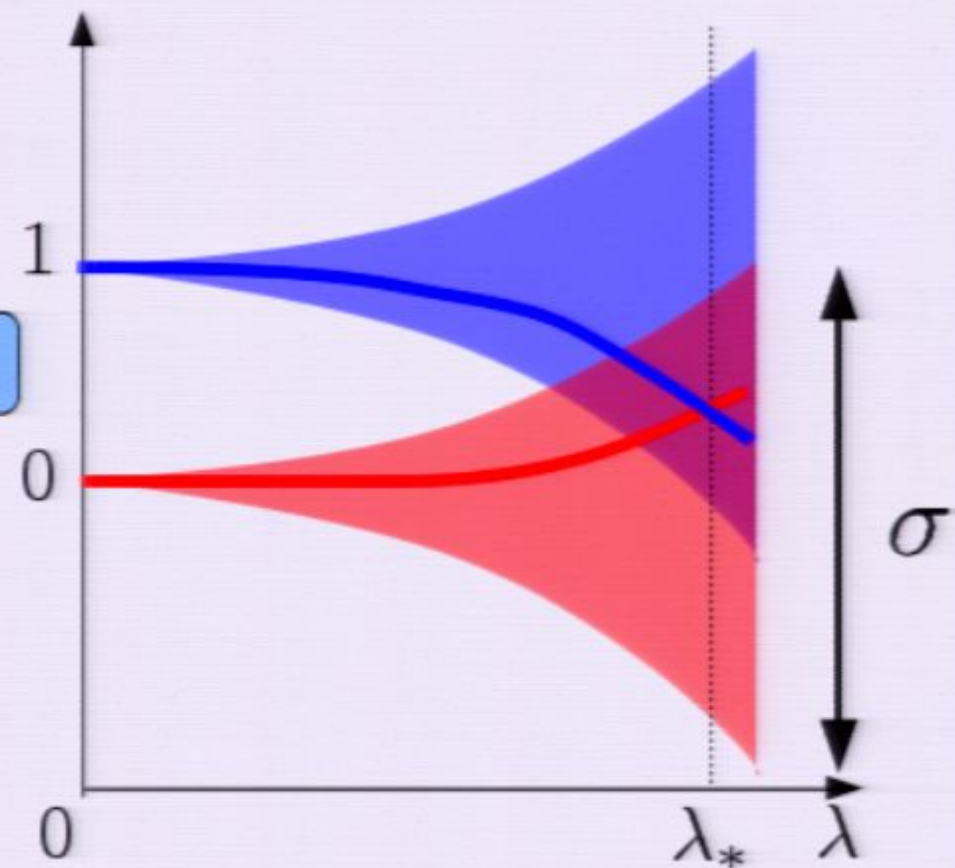
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Random instances $\Rightarrow$ random slopes!



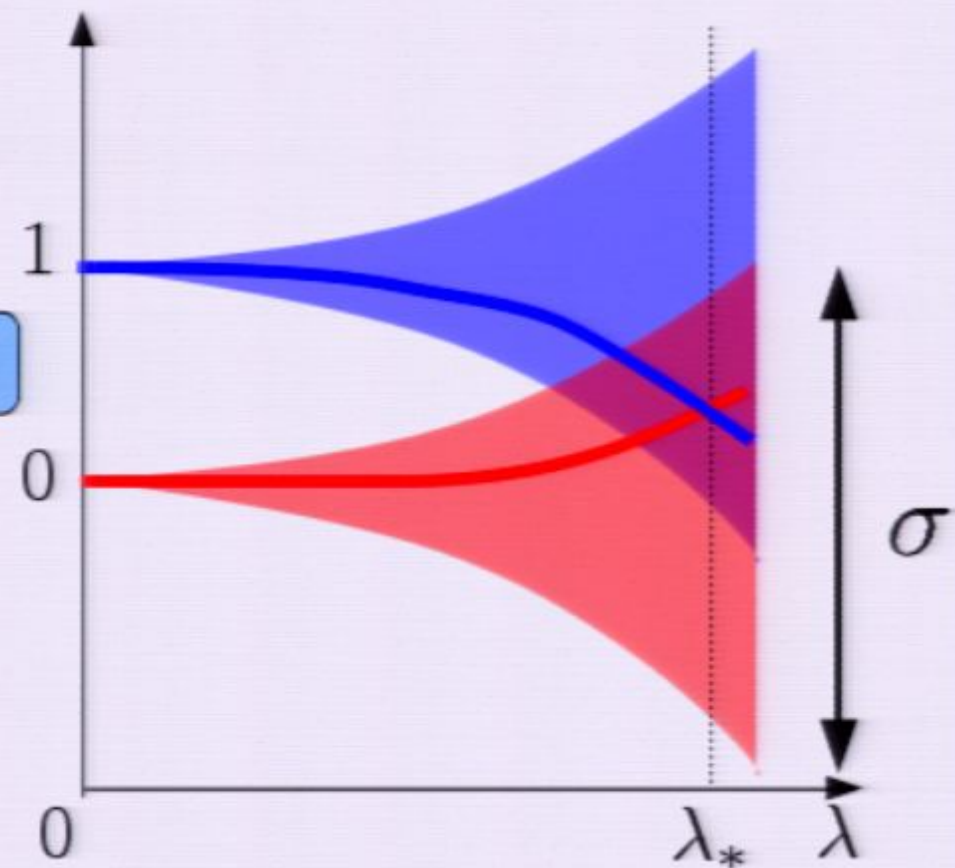
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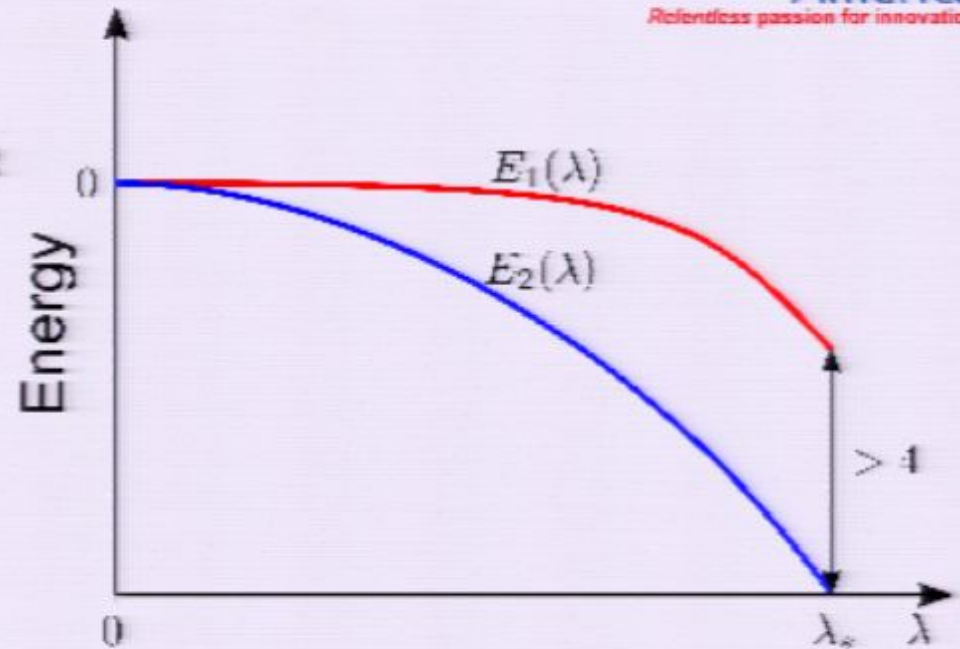
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Suppose

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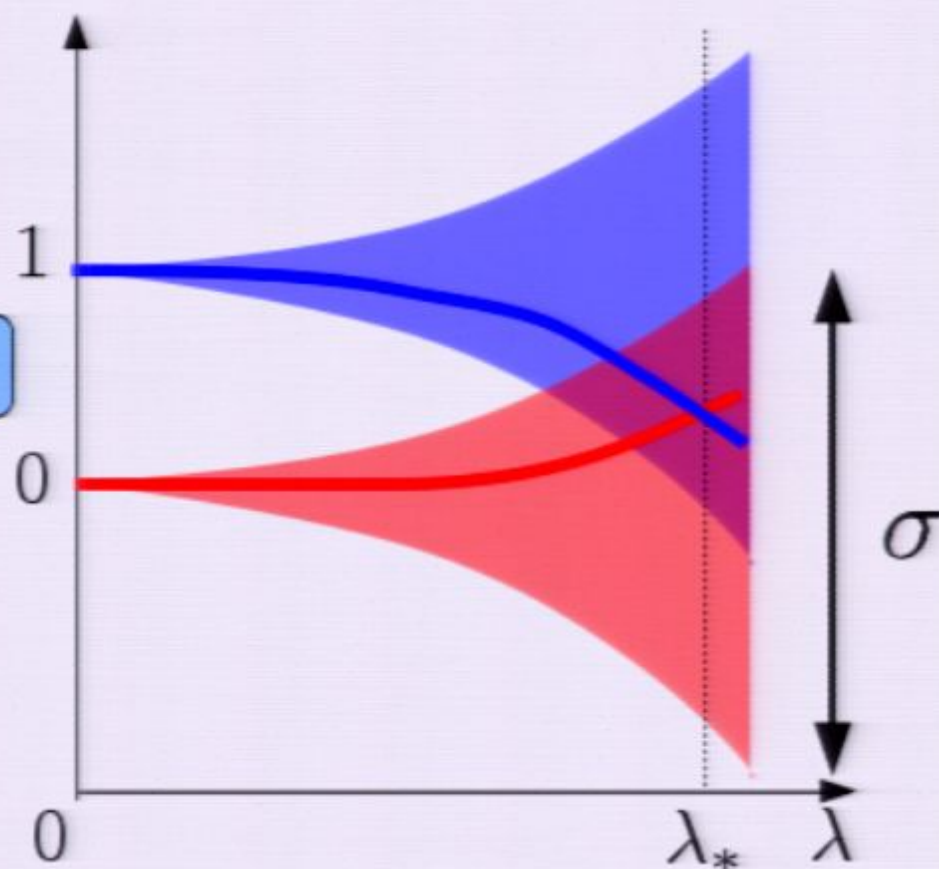
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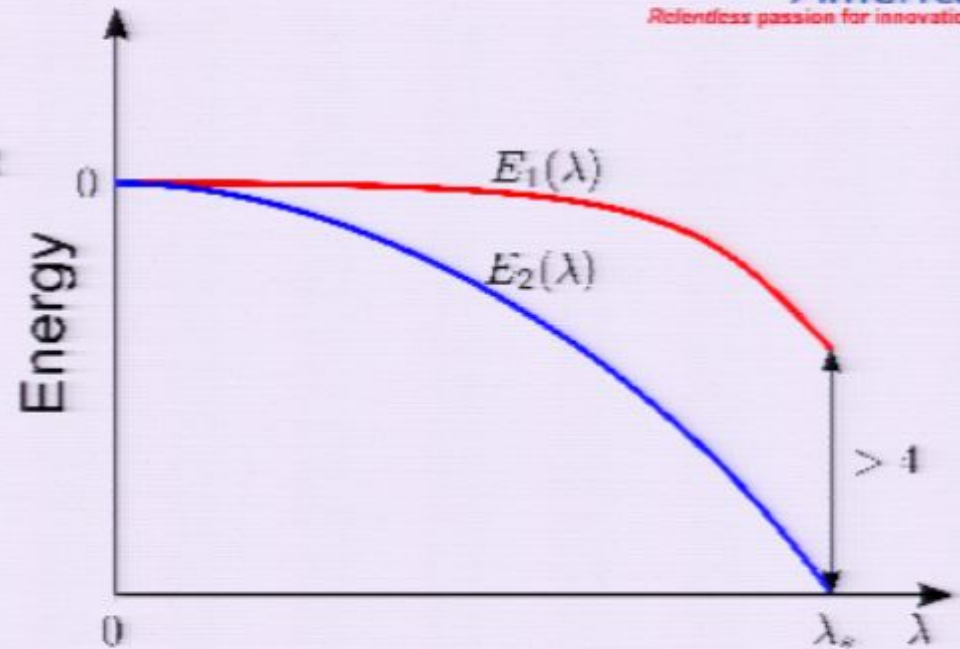
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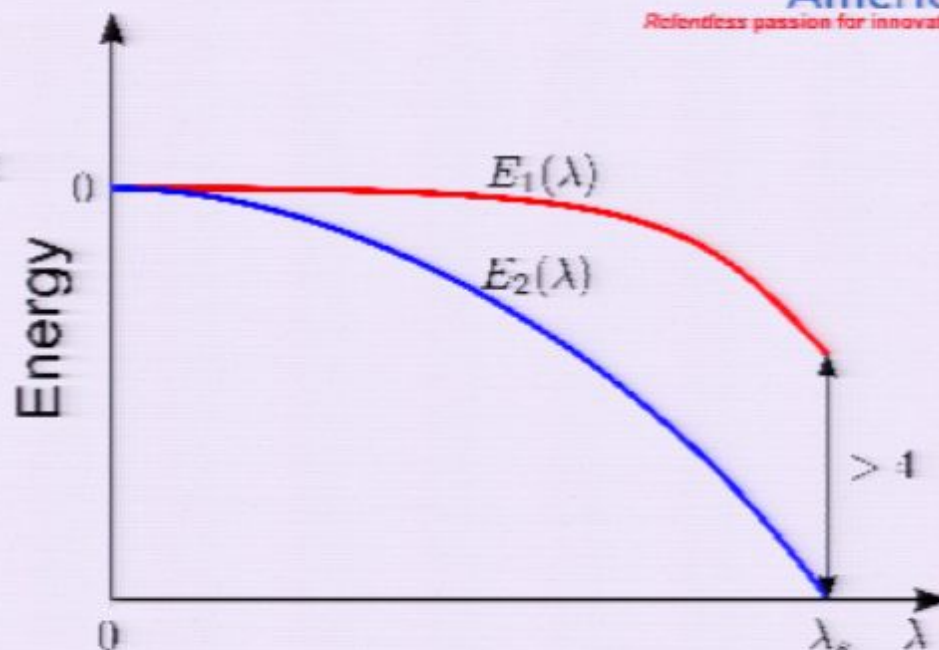
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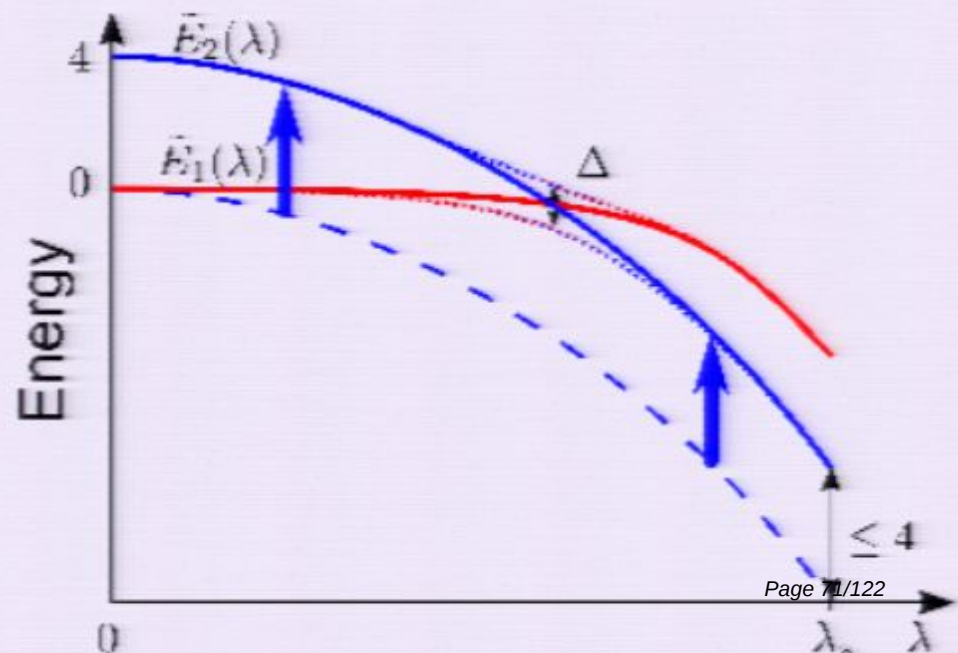
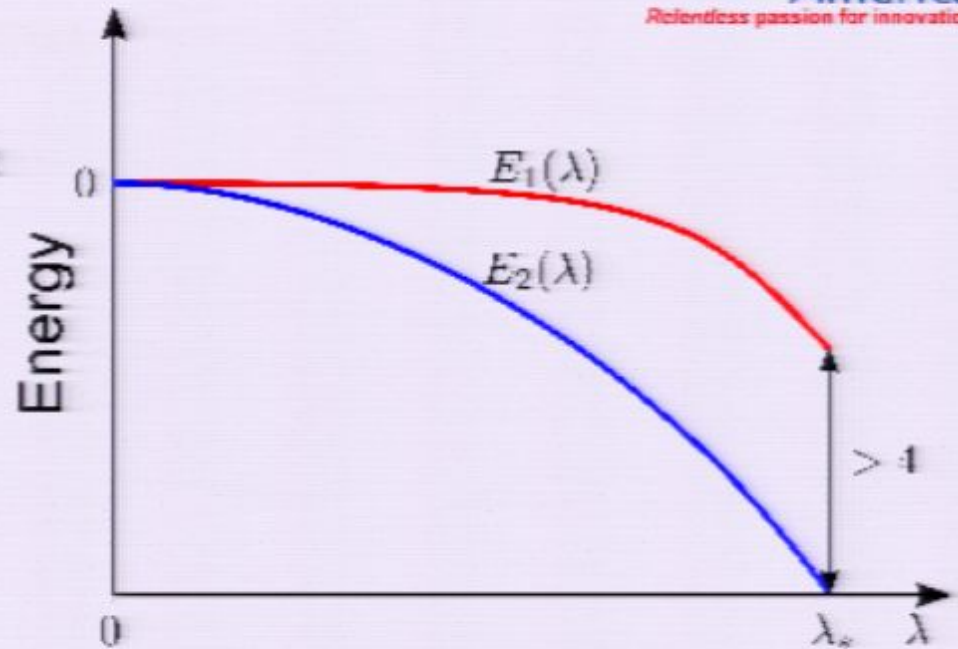
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We compute $E_{1,2}(\lambda)$ by perturbation theory

$$E_{\vec{x}}(\lambda) = E_{\vec{x}}(0) + \sum_{m=1}^{\infty} \lambda^{2m} F_{\vec{x}}^{(m)}$$

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Proof based on statistical
properties of random instances

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Numerical simulations

- We generated EC3 random instances with >2 solutions
- then computed $E_1(\lambda) - E_2(\lambda)$ by order 4 perturbation theory

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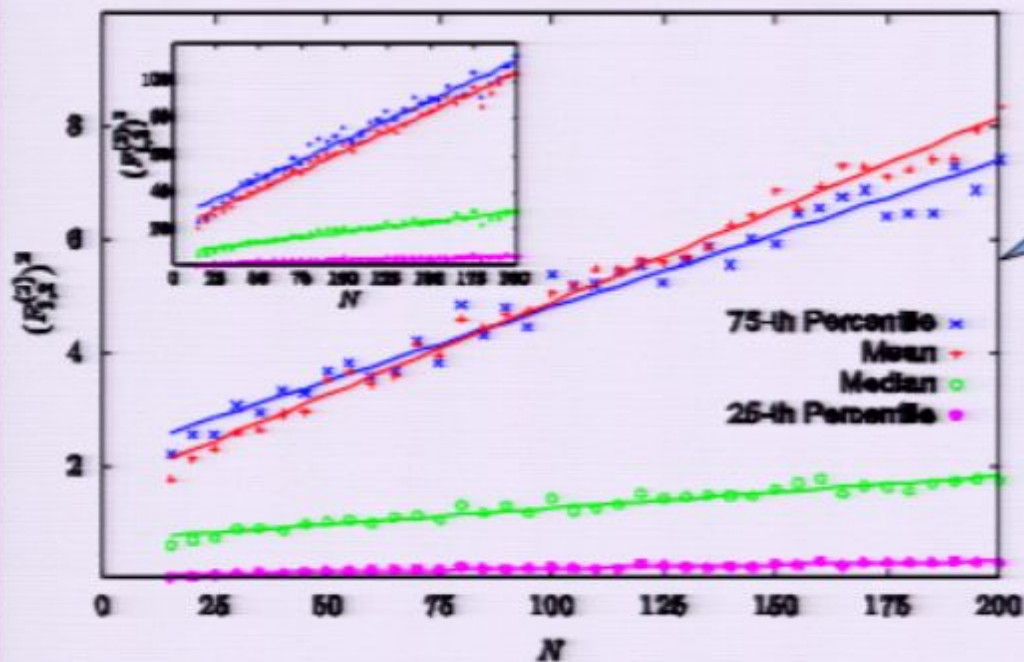
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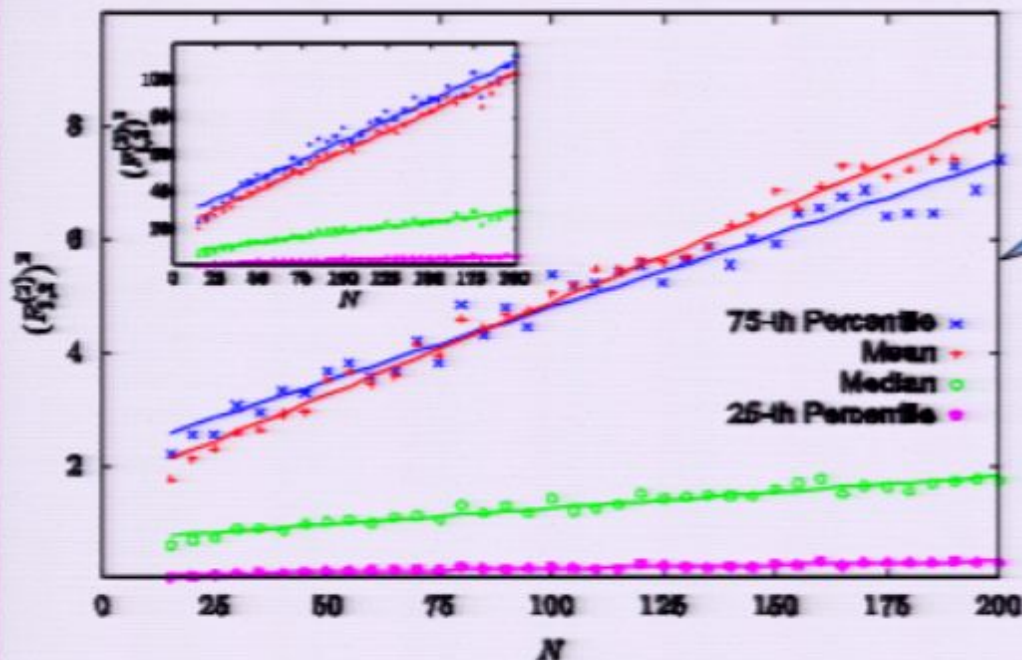


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We have $E_1(\lambda) - E_2(\lambda) > 4$ for $\lambda > \sqrt{2}(CN)^{-1/8}$

How small is the gap?

We show that up to leading order in perturbation theory:

$$\Delta < (2\lambda_*)^n$$

Proof by reduction to the “Agree” problem:
2-bit clauses $(x_{i_C} = x_{j_C})$

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Anderson localization theory

$\Rightarrow$ Perturbation theory valid as long as states are localized

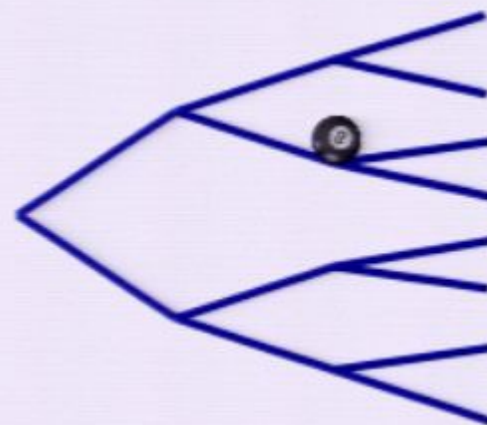
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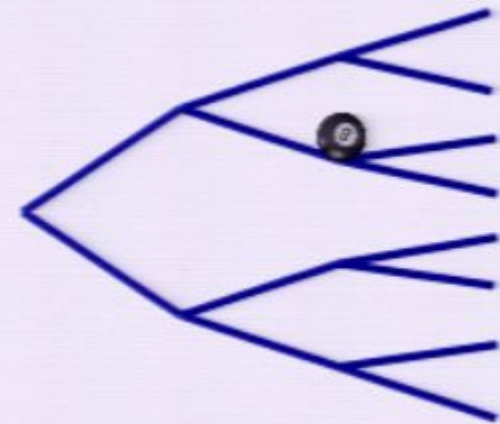
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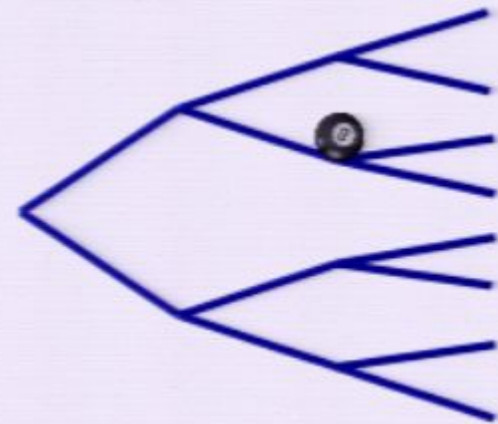
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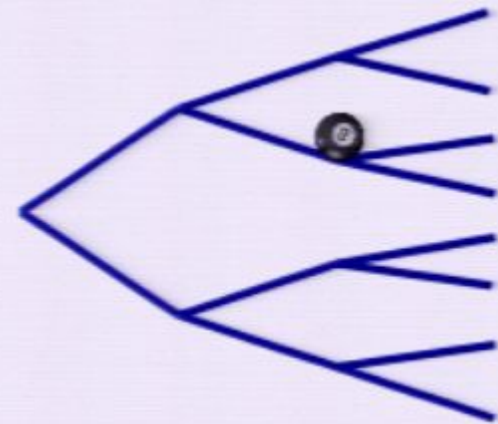
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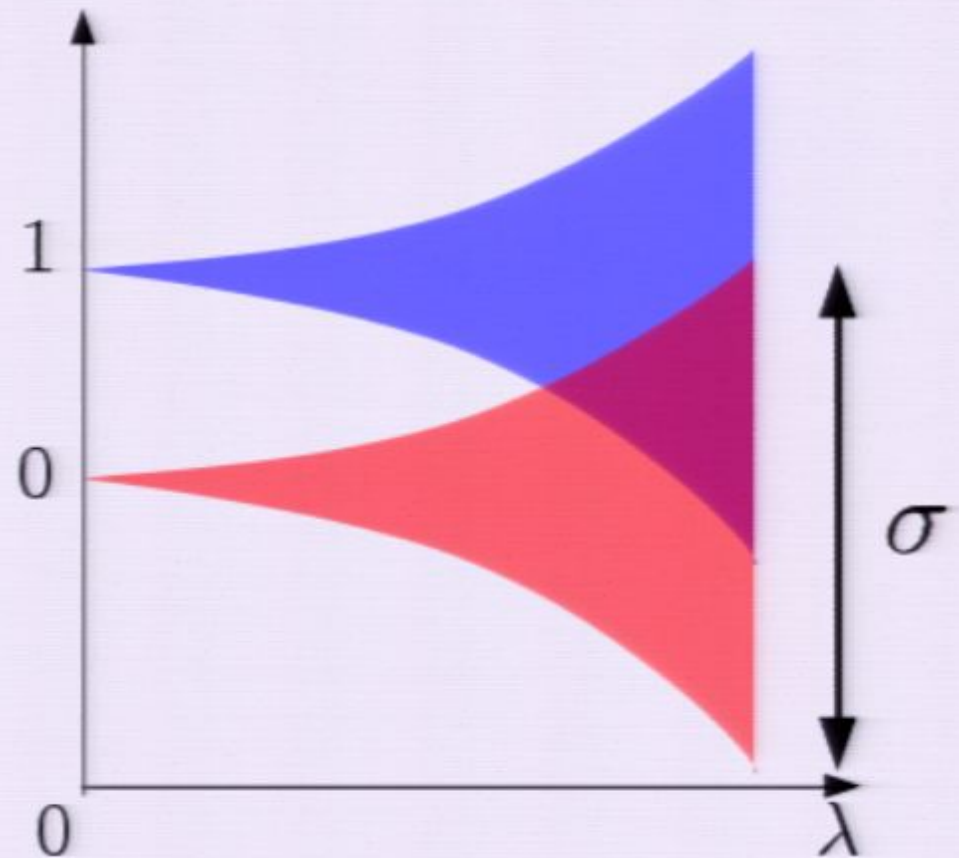
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Our estimation: $\sigma \sim \sqrt{N} \lambda^4$



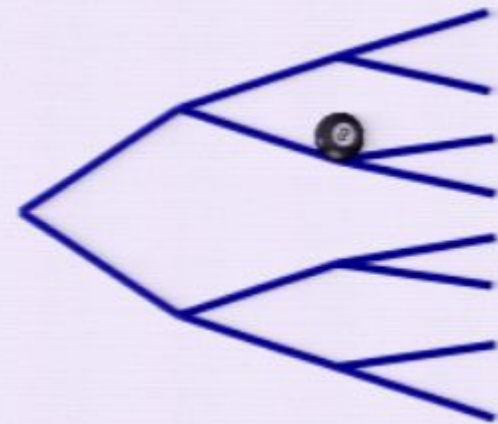
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• Ground state is degenerate

$$S_1 \sim e^{\eta N} \Rightarrow \sigma_1 \sim \lambda^4$$

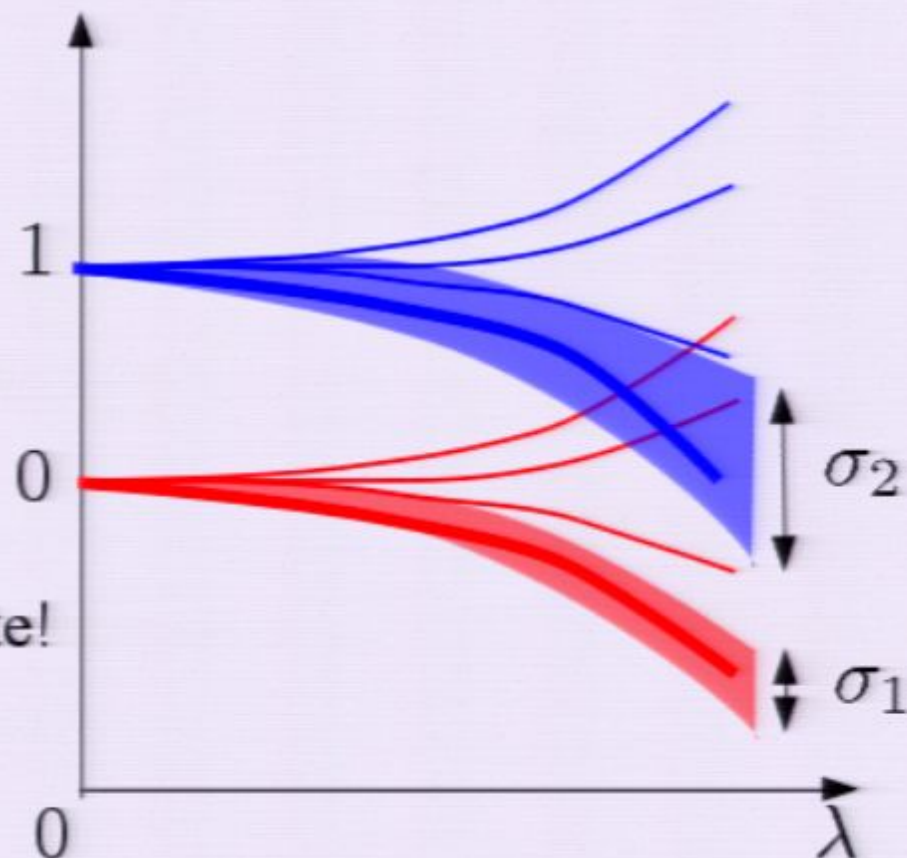
[Knysh-Smelyanskiy'10]

BUT

• First excited state is more degenerate!

$$S_2 \sim N e^{\eta N} \Rightarrow \sigma_2 \sim \frac{\log N}{\sqrt{\eta}} \lambda^4$$

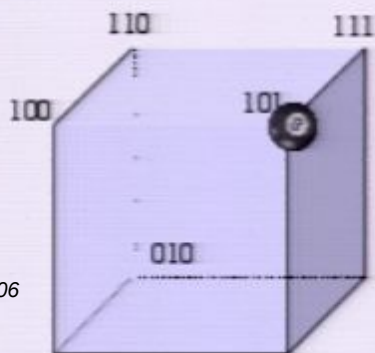
• Also: $\eta \rightarrow 0$ as $\alpha \rightarrow \alpha_s$



$\Rightarrow$ Effect of degeneracy only appears for large N

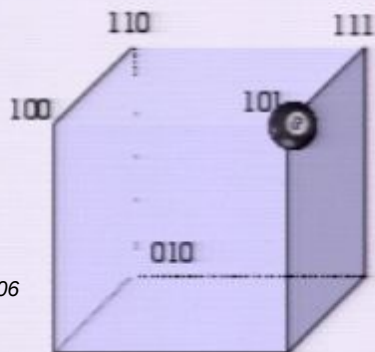
Conclusion

- Anderson localization causes exponentially small gaps in adiabatic quantum optimization



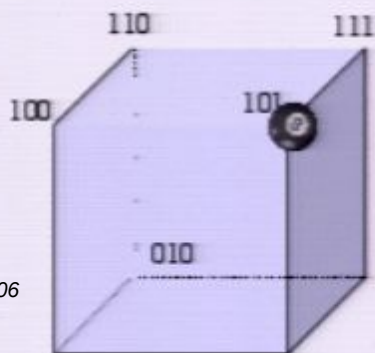
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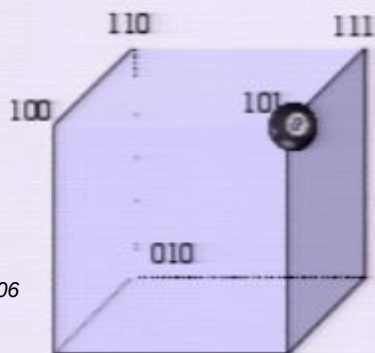
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- Important assumption: Localization on the hypercube



$\Rightarrow$ should be studied more closely

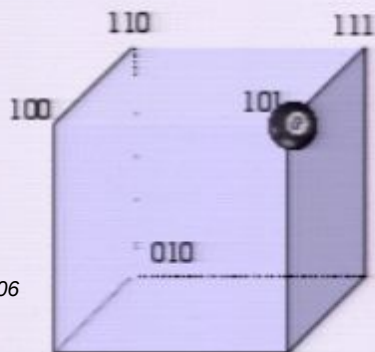
Thank you!

www.nec-labs.com

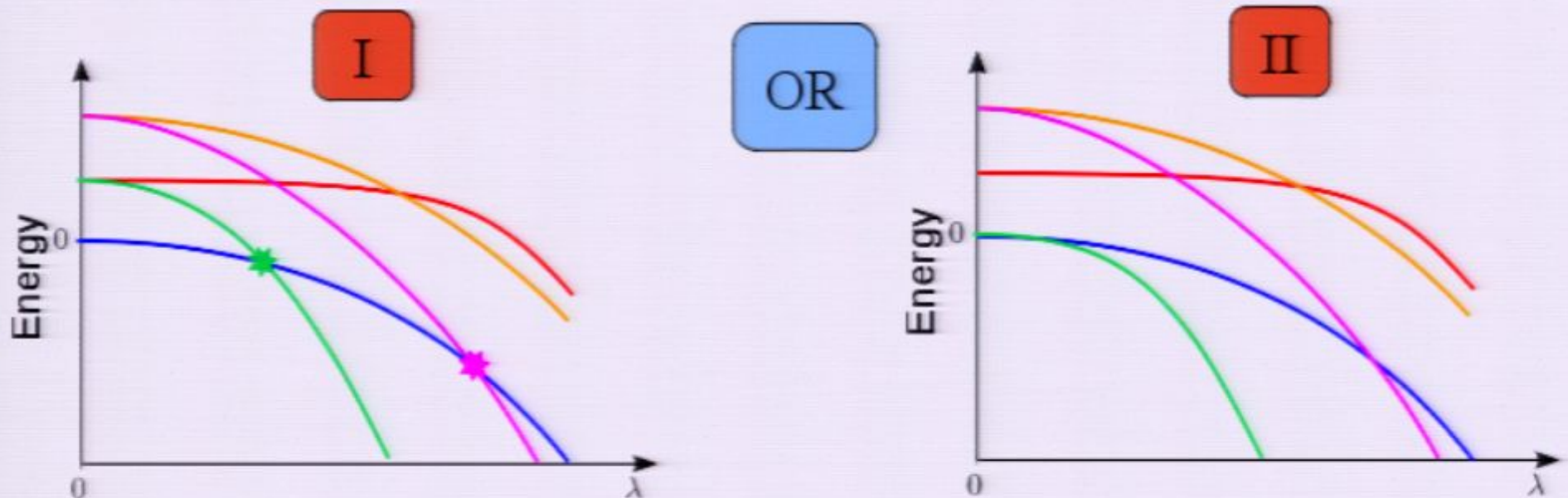


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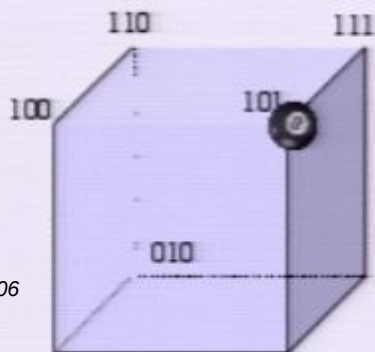
Effect of path change



- Idea: Pick random $H(s)$ (“path change”) to obtain case II [Farhi et al. '09]
- Avoid 1 crossing: $Pr[\text{“II”}] = \text{constant}$
- Avoid poly # of crossings: $Pr[\text{“II”}] = 1/\text{poly}$
- Estimated # of crossings: $\exp(N/\log^7 N)$

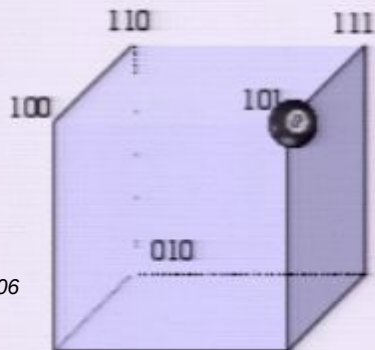
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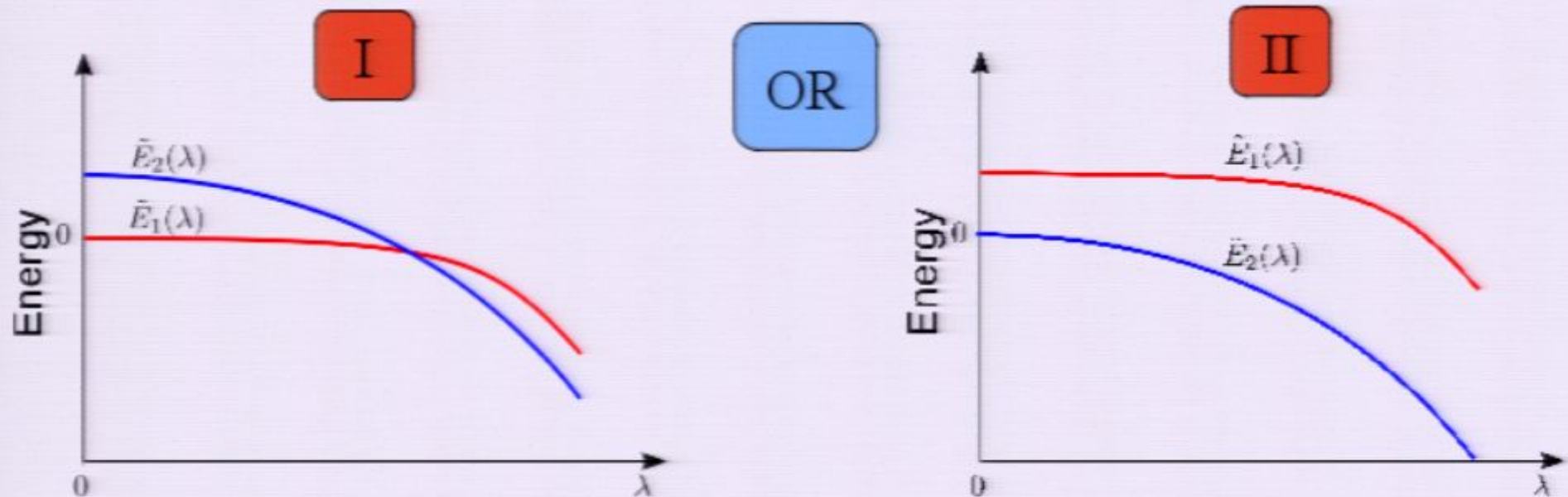


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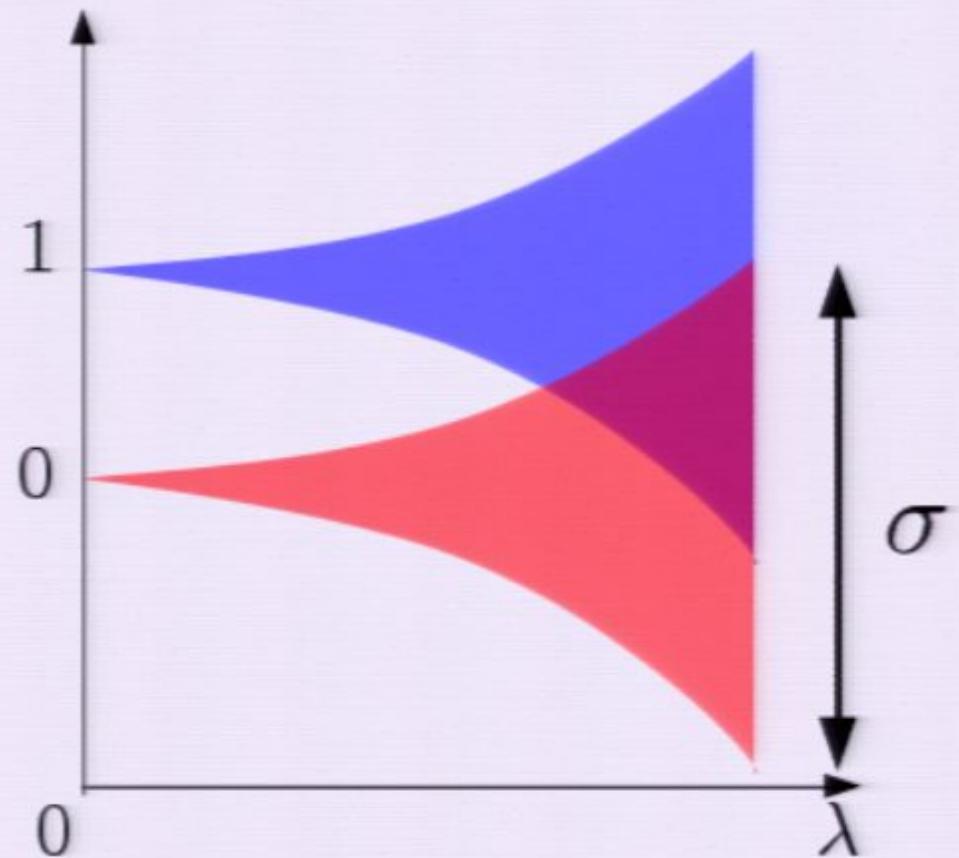
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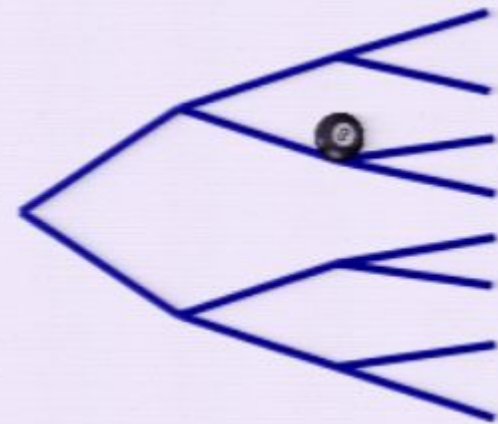
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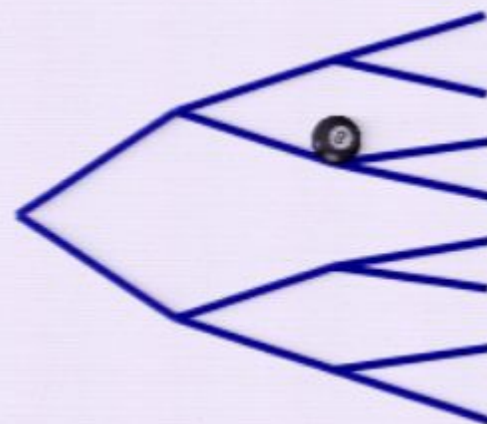
Can we trust perturbation theory?

Anderson localization theory

$\Rightarrow$ Perturbation theory valid as long as states are localized

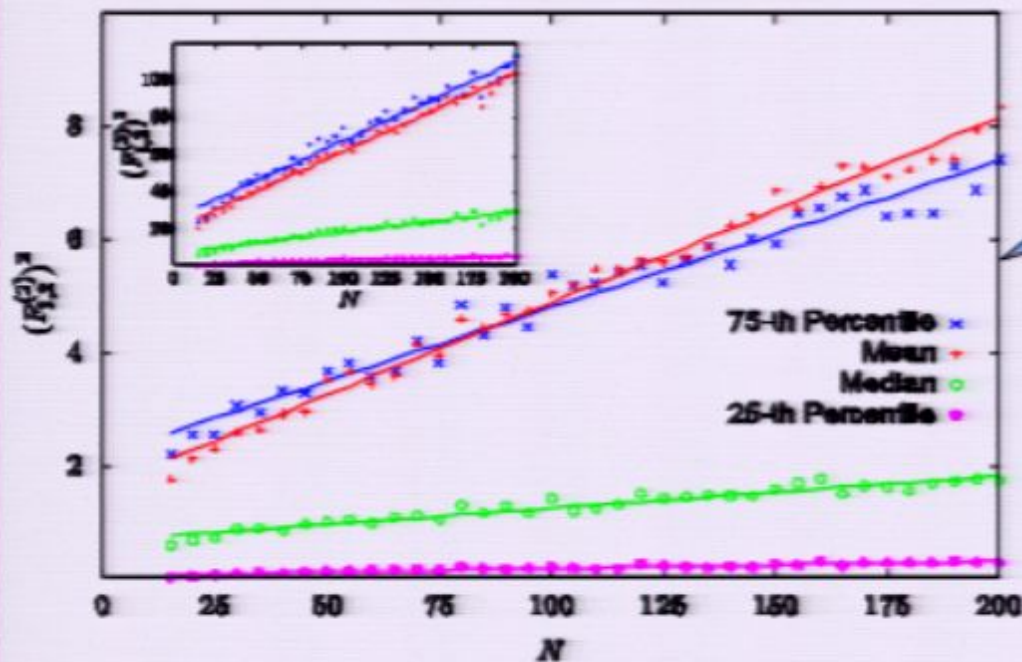
Cayley tree with branching number K :

$$\lambda_c = \Theta \left(\frac{E}{K \log K} \right)$$



How small is the gap?

- We generated EC3 random instances with >2 solutions
- then computed $E_1(\lambda) - E_2(\lambda)$ by order 4 perturbation theory



Each data point computed from 2500 instances

$$(E_1(\lambda) - E_2(\lambda))^2 \approx CN\lambda^8$$

We have $E_1(\lambda) - E_2(\lambda) > 4$ for $\lambda > \sqrt{2}(CN)^{-1/8}$

How small is the gap?

We show that up to leading order in perturbation theory:

$$\Delta < (2\lambda_*)^n$$

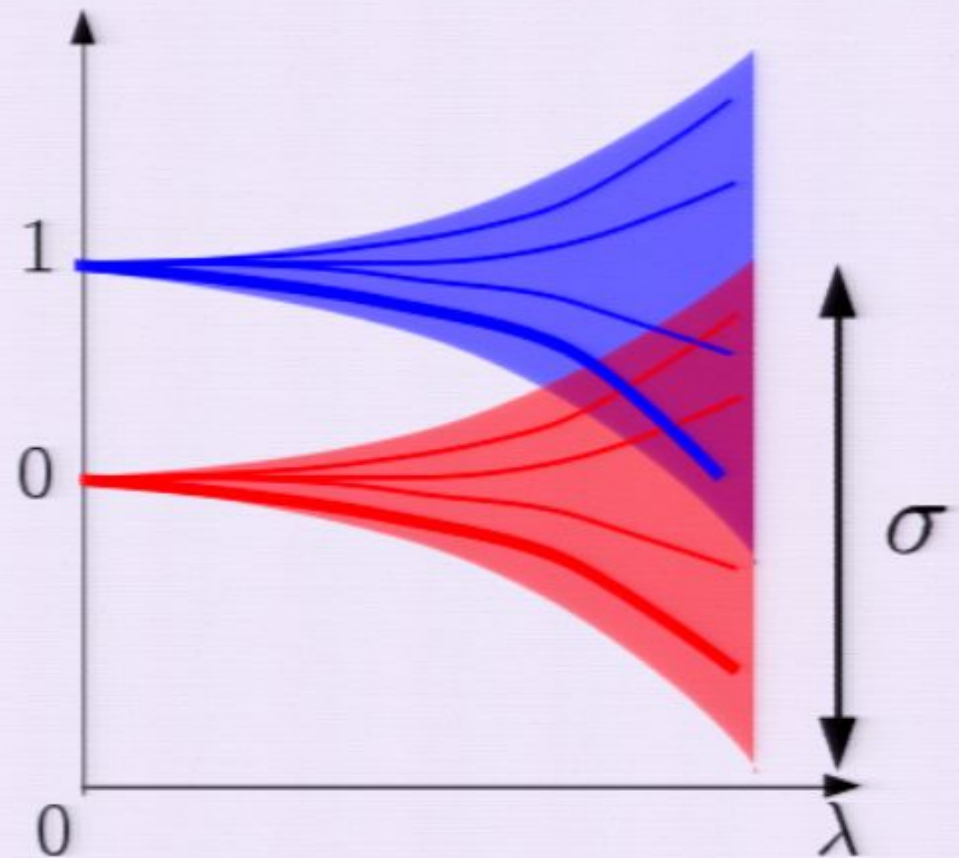
Since: 1) level crossings appear at $\lambda_* = O(N^{-1/8})$

2) typical distance between solutions is $n = \Theta(N)$

Degeneracy of the ground state

• Our estimation: $\sigma \sim \sqrt{N}\lambda^4$

• Ground state is degenerate



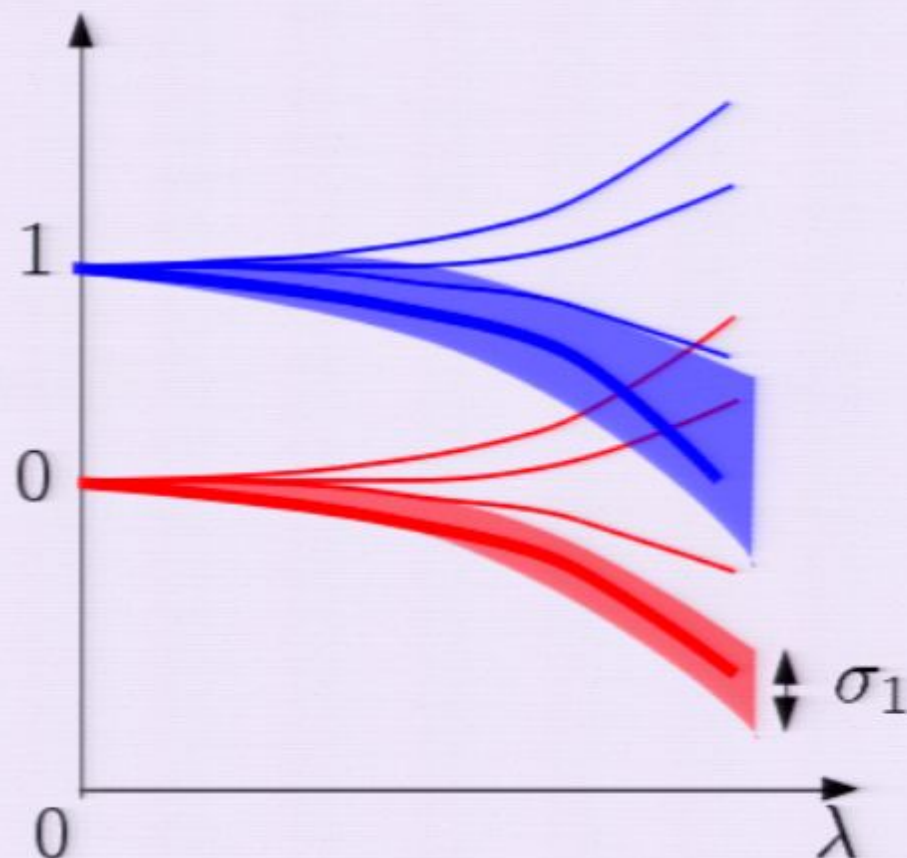
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$$S_1 \sim e^{\eta N} \Rightarrow \sigma_1 \sim \lambda^4$$

[Knysh-Smelyanskiy'10]



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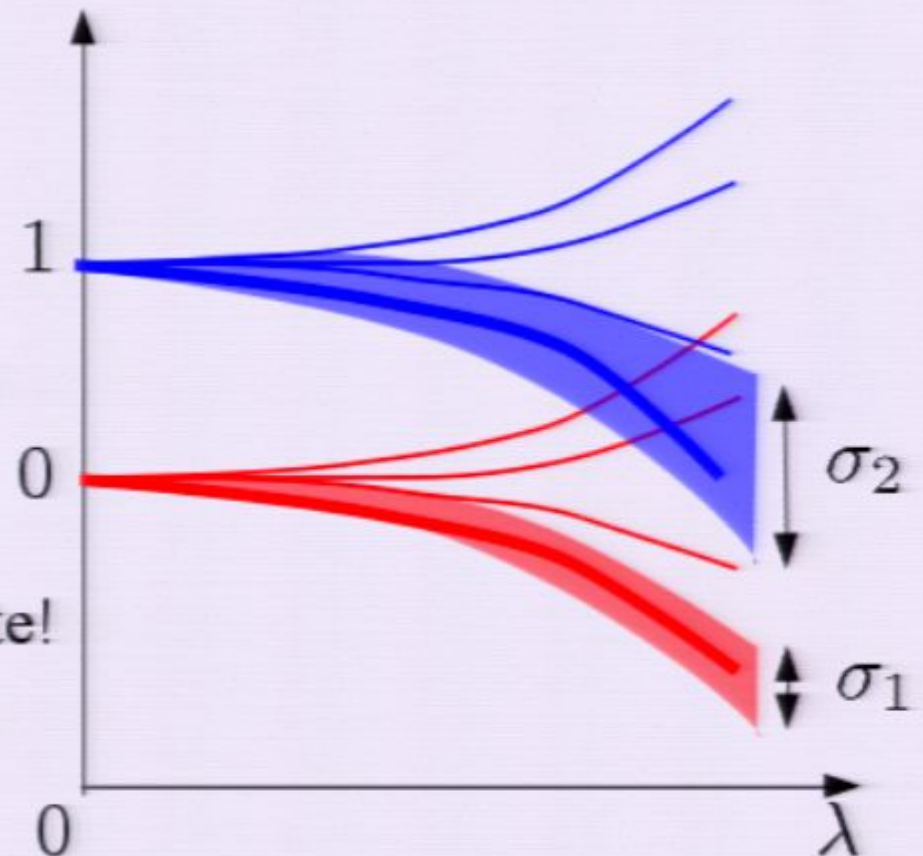
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[Knysh-Smelyanskiy'10]

BUT

• First excited state is more degenerate!

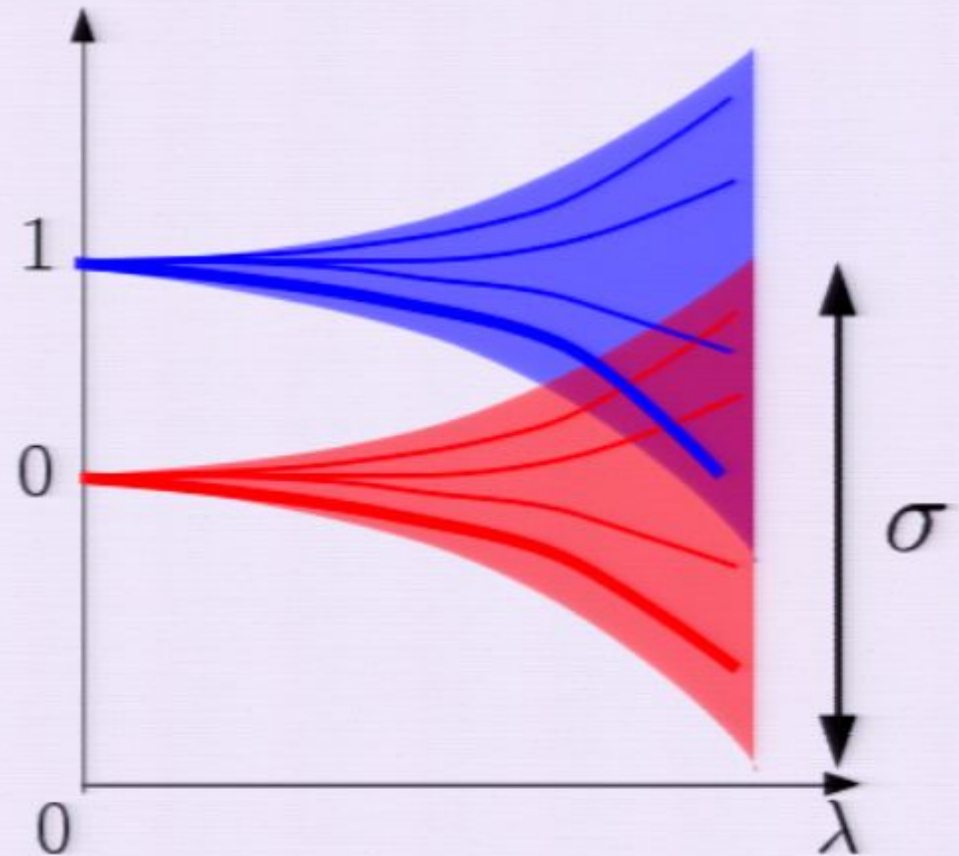
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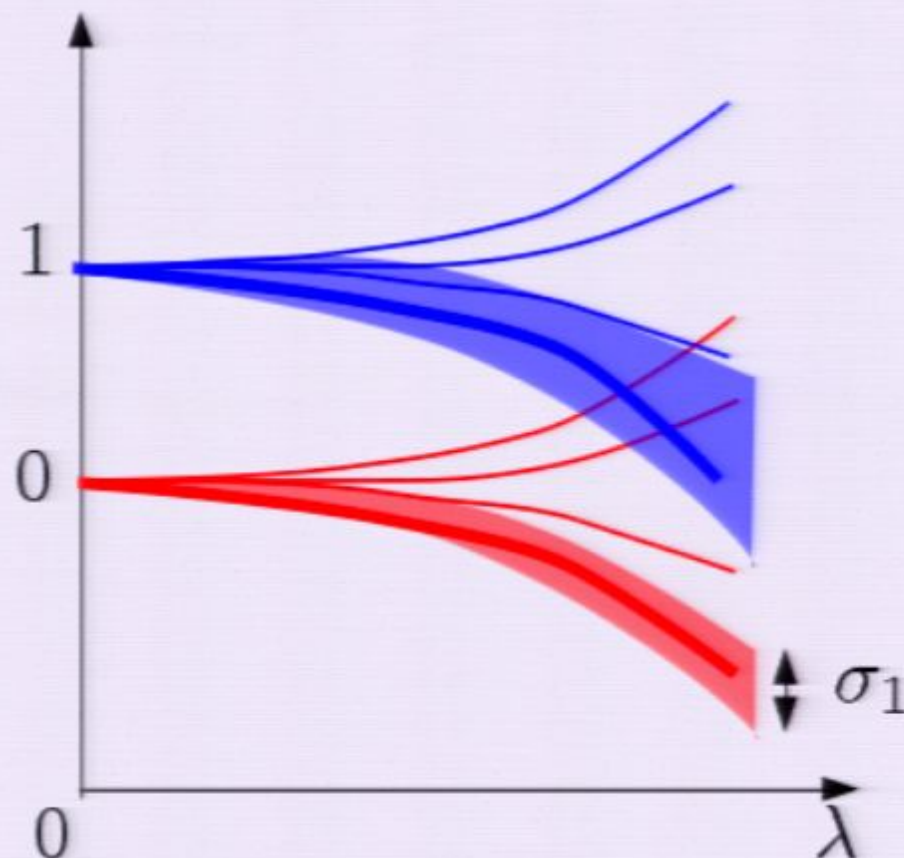
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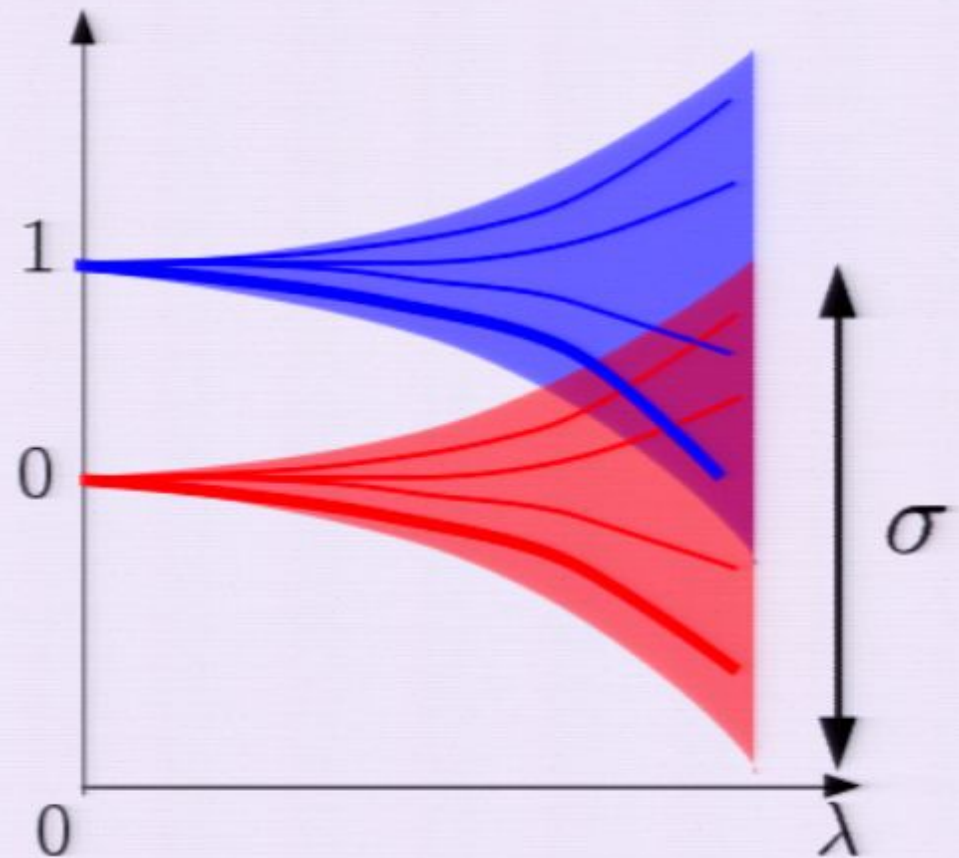
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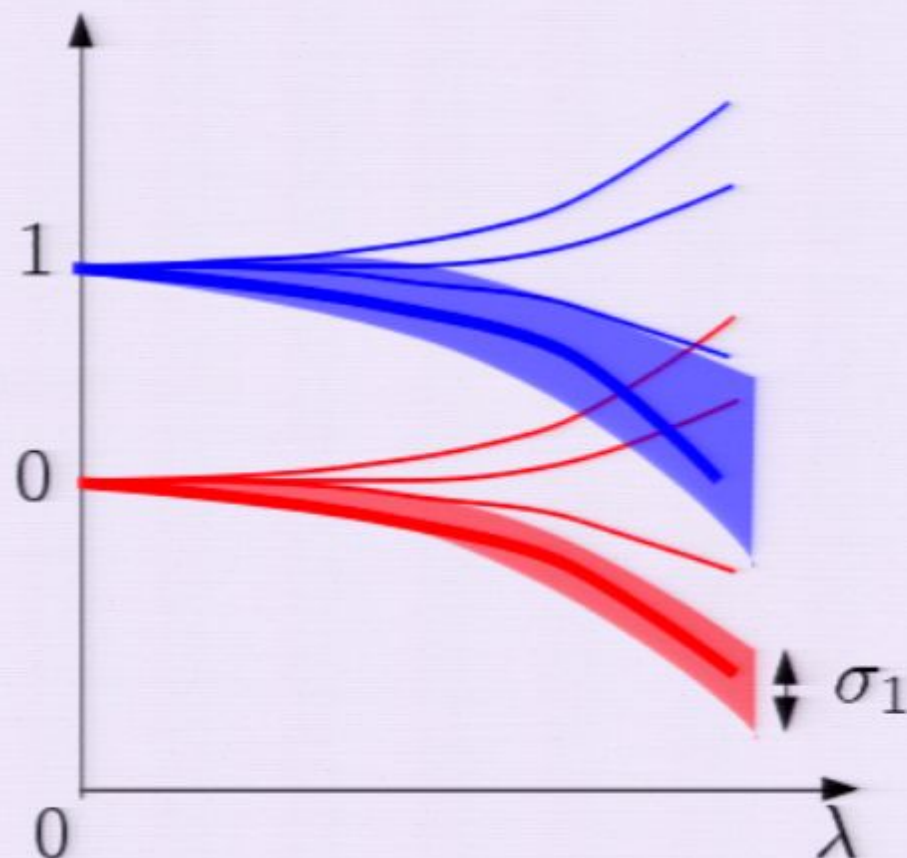
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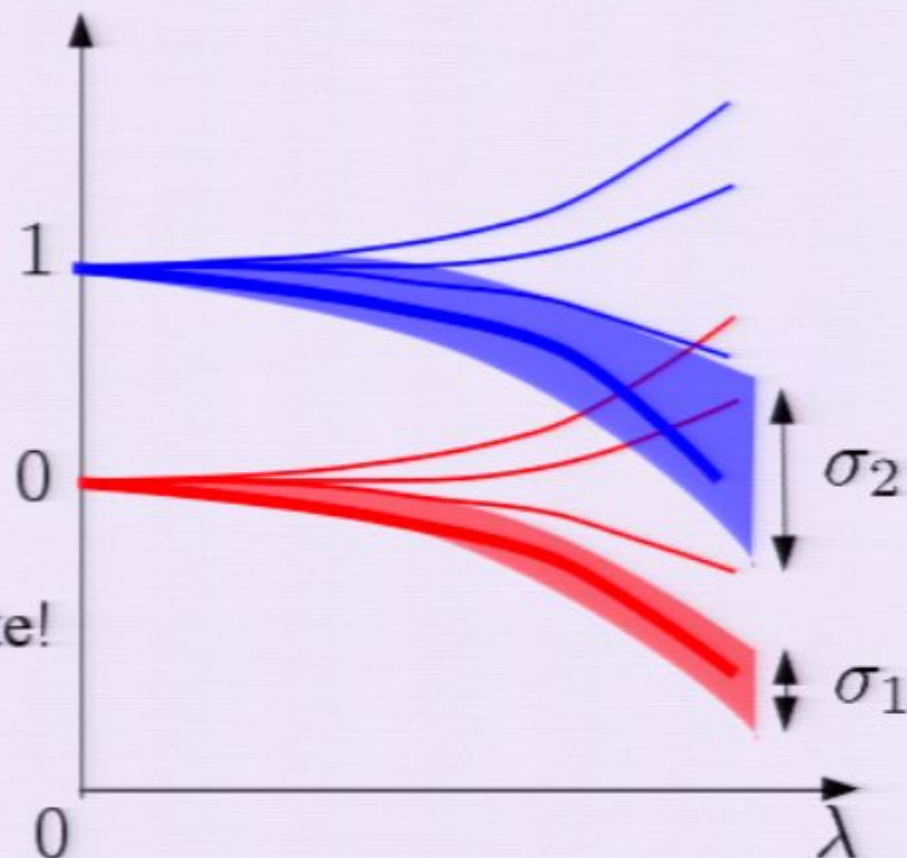
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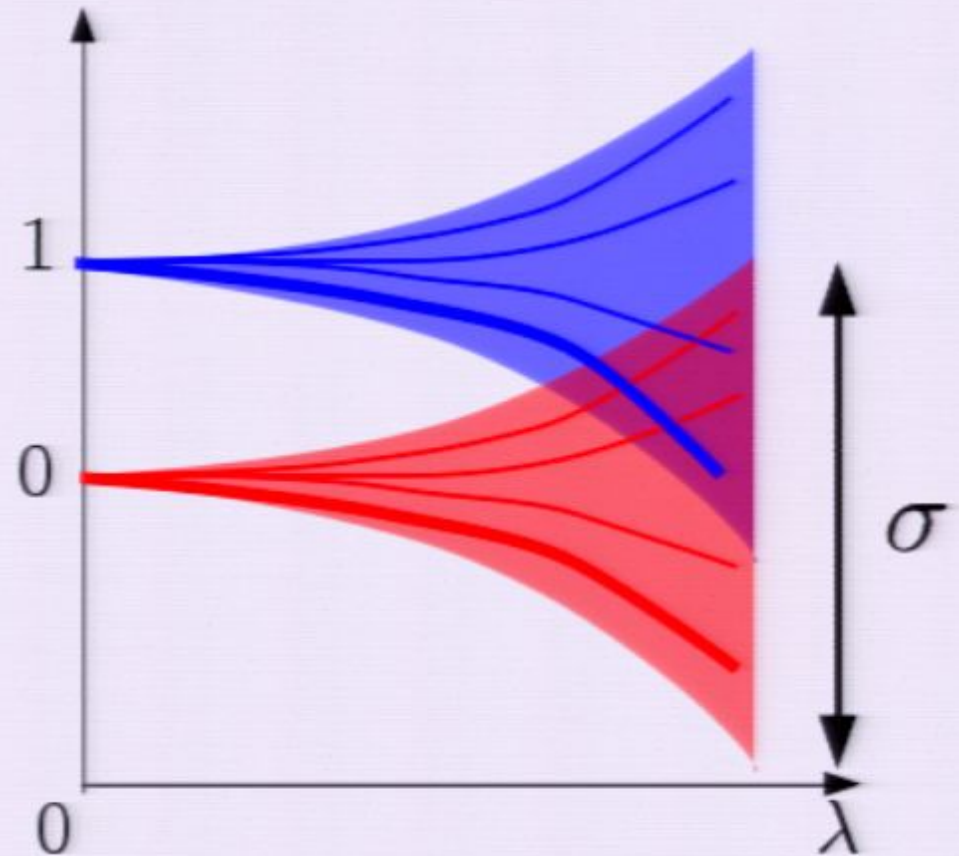
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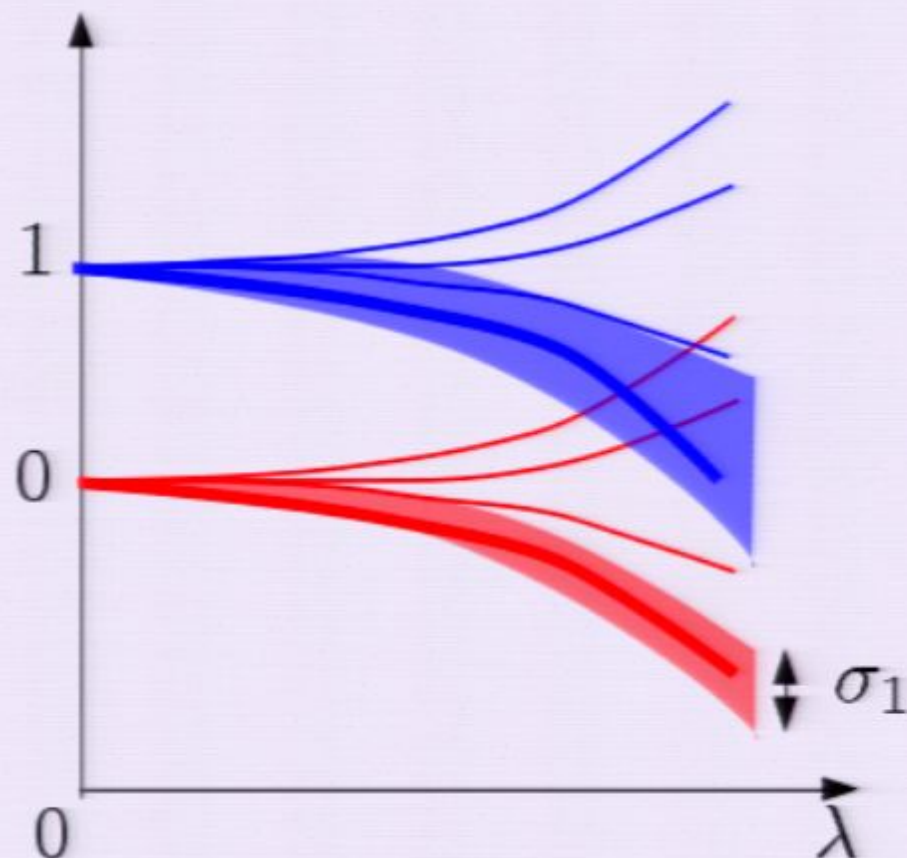
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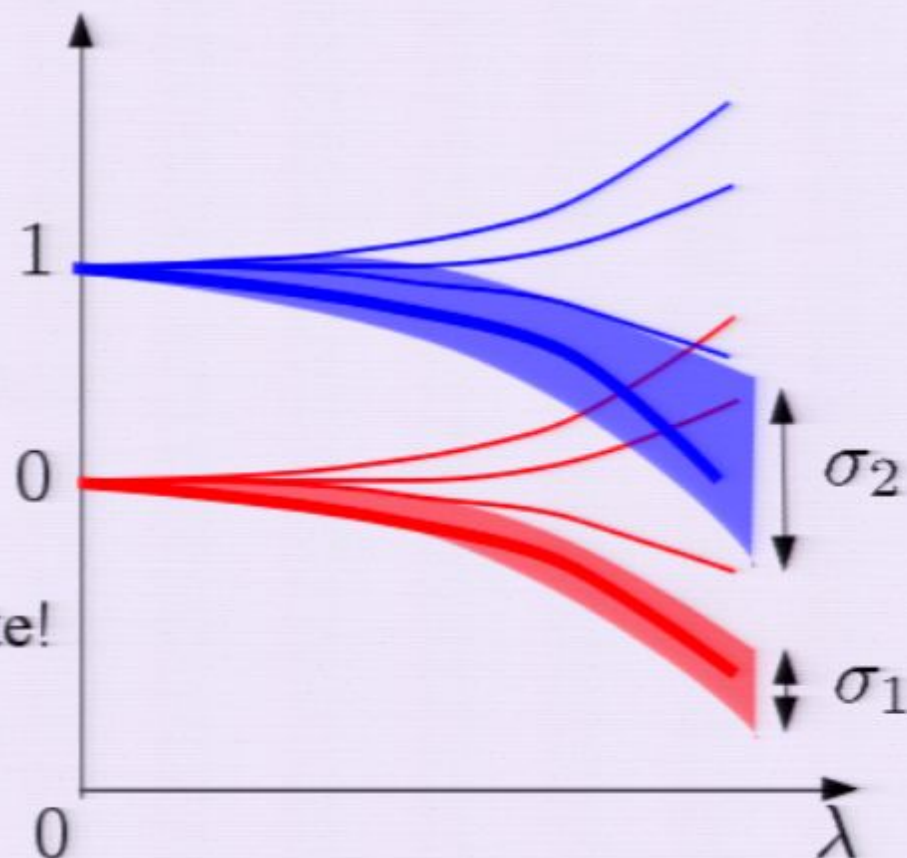
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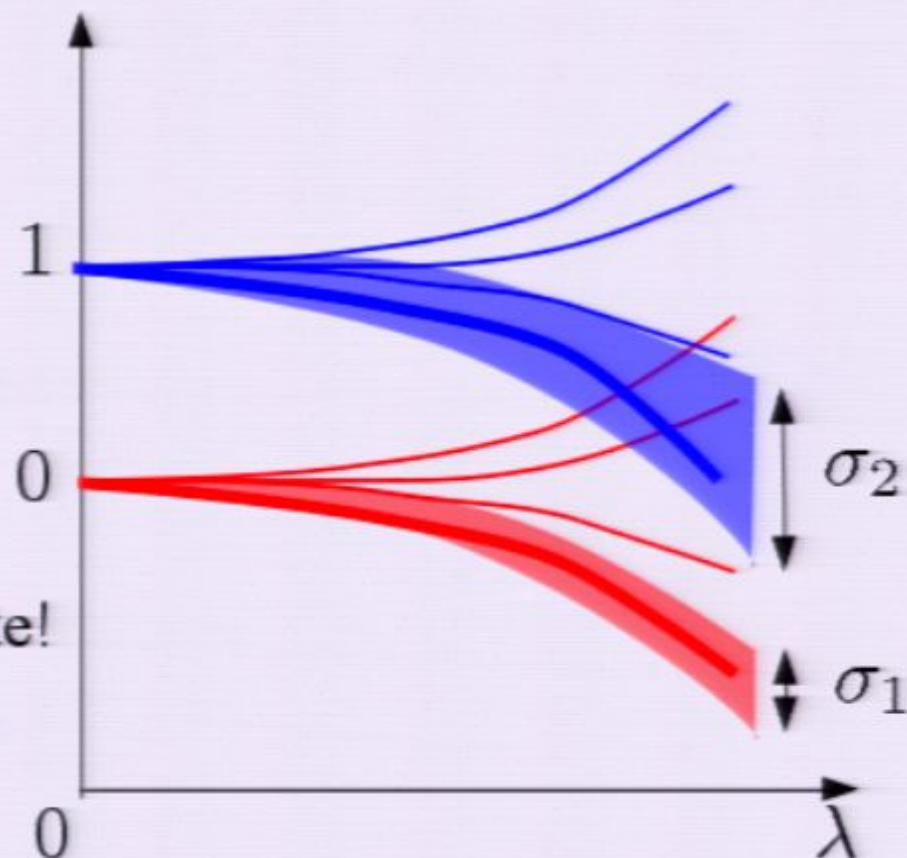
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• Also: $\eta \rightarrow 0$ as $\alpha \rightarrow \alpha_s$

$\Rightarrow$ Effect of degeneracy only appears for large N