



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result



# Statistics of Geometric Phase Curvature

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(i) J. Phys A, 51, 475101, (2018)  
(ii) J. Stat. Phys., Springer Nature, (2019)

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# Geometric Curvature



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Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

- Geometric phase defined for a closed loop when the Hamiltonian goes around a closed loop  $\mathbf{R}(T) = \mathbf{R}(0)$  in parameter space,
- circuit dependent in general
- need for a closed loop usually leads to technical issues
- often better to use "local phase": **geometric curvature**

$$\gamma_n(T) = i \oint_{\mathbf{R}(0)}^{\mathbf{R}(T)} \langle n, \mathbf{R}(\tau) | \nabla_{\mathbf{R}} | n, \mathbf{R}(\tau) \rangle \cdot d\mathbf{R} = \iint \mathbf{C}_n \cdot d\mathbf{S}$$

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**Geometric curvature**      $\mathbf{C}_n \equiv \nabla_{\mathbf{R}} \times i \langle n, \mathbf{R} | \nabla_{\mathbf{R}} | n, \mathbf{R} \rangle$

- Geometric phase for a loop  $\rightarrow$  flux of the Geometric curvature

# Curvature is better than phase !



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Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

- Geometric phase always associated with a closed path  
Geometric curvature a local quantity  
provides a local description of the geometric properties of the parametric space.
- both gauge-invariant, can act as observables
- Geometric curvature: a geometric magnetic field  
Just as a magnetic field can affect the dynamic of an electron, curvature directly affects the dynamic of parameters (adiabatic).  
In this sense, curvature is more fundamental quantity than phase.
- no closed loop needed for curvature analysis
- curvature has many applications: example: essential in a proper description of thr Bloch electrons which has various effects on transport and thermodynamics properties

$$C_n(\mathbf{R}) = \text{Im} \nabla_{\mathbf{R}} \langle \mathbf{n}, \mathbf{R} | \times \nabla_{\mathbf{R}} | \mathbf{n}, \mathbf{R} \rangle$$

# in terms of eigenvalues and eigenfunctions



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$C_n(\mathbf{R}) = 2 \text{Im} \sum_{\mathbf{m} \neq \mathbf{n}} \frac{\langle \mathbf{n} | \nabla_{\mathbf{R}} \mathbf{H} | \mathbf{m} \rangle \times \langle \mathbf{m} | \nabla_{\mathbf{R}} \mathbf{H} | \mathbf{n} \rangle}{(E_n - E_m)^2}$$

**Hamiltonian  $H(\mathbf{R})$  in 3-dimensional parameter space  $\mathbf{R} = (x, y, z)$**

$$\mathbf{C}_n = (C_{nx}, C_{ny}, C_{nz}) \quad (\text{in vector form})$$

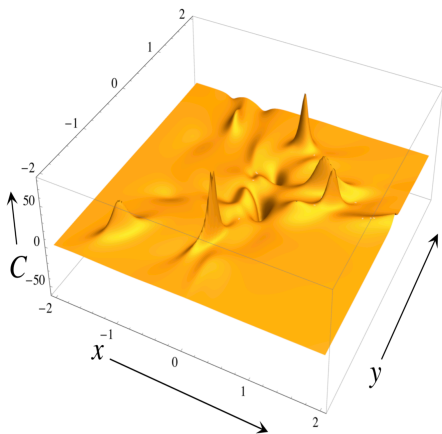
$$C_{nz}(x, y) = 2 \text{Im} \sum_{\mathbf{m} \neq \mathbf{n}} \frac{\langle \mathbf{n} | \partial_x \mathbf{H} | \mathbf{m} \rangle \langle \mathbf{m} | \partial_y \mathbf{H} | \mathbf{n} \rangle}{(E_n - E_m)^2}$$

**what is geometric curvature for a complex system?**

# Curvature of a typical random state:

$$H(x, y) = H_0 + xH_x + yH_y$$

for the state  $n = 10$  of a  $20 \times 20$  Hamiltonian with  $H_0, H_x, H_y$  sampled from the GUE.



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Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

# Role of Complexity



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

- Lack of detailed knowledge about interactions manifests itself by randomization (partial / full) of various operators of the system.
- Determination of some/ all matrix elements  $\Rightarrow$  within an uncertainty  $\Rightarrow$  matrix elements are at best described by a probability distribution.
- Uncertainty associated with various matrix element can be of different types  $\Rightarrow$  matrix elements can have different distributions.
- Hamiltonian  $\Rightarrow$  a random matrix (**some/ all matrix elements randomly distributed**)
- Distributions of the eigenfunctions and eigenvalues
- What is the distribution of geometric curvature  $\mathbf{P}_c(\mathbf{C})$  ?

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# distribution of curvature



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

Complexity leads to sample to sample fluctuations in eigenvalues and eigenfunctions  $\rightarrow$  fluctuations of geometric curvature

- **distribution of curvature components ( $C_x, C_y, C_z$ ) of an arbitrary eigenstate**

$$P(C_x, C_y, C_z) = \langle \delta(C_x - C_{nx}) \delta(C_y - C_{ny}) \delta(C_z - C_{nz}) \rangle$$

with

$$C_{nz} = 2 \operatorname{Im} \sum_{m \neq n} \frac{\langle n | \partial_x H | m \rangle \langle m | \partial_y H | n \rangle}{(E_n - E_m)^2}$$

similarly

$$C_{nx}, C_{ny}$$

**needed: distribution of the eigenvalues and eigenfunctions of  $H$**

# begin with a simple question



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

Hamiltonian  $H(\mathbf{R})$  in 2-dimensional parameter space  $\mathbf{R} = (x, y)$

For 
$$\mathbf{H}(x, y) = \mathbf{H}_0 + x \mathbf{H}_x + y \mathbf{H}_y$$

$$\mathbf{C} \equiv \mathbf{C}_n(x, y) = 2 \operatorname{Im} \sum_{m \neq n} \frac{\langle n | \mathbf{H}_x | m \rangle \langle m | \mathbf{H}_y | n \rangle}{(\mathbf{E}_n - \mathbf{E}_m)^2}$$

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# 2 × 2 Hamiltonian



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

- only two states with curvature  $C_1, C_2$

$$C_1 = -C_2 = 2 \operatorname{Im} \frac{\langle 1 | H_x | 2 \rangle \langle 2 | H_y | 1 \rangle}{(E_1 - E_2)^2} = \frac{-i(F - F^*)}{S^2}$$

- $E_2 - E_1 \equiv S$ : spacing between eigenvalues  $E_1, E_2$
- Perturbation  $H_x, H_y$  measured in eigenfunction basis of  $H$**   
 $F \equiv F_{12} \equiv \langle 1 | H_x | 2 \rangle \langle 2 | H_y | 1 \rangle = F_{21}^*$
- Desired Probability distribution**

$$\begin{aligned} P_c(C) &= \langle \delta[C - C_1] \rangle \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} dt \exp[iCt] \langle \exp[-iC_1 t] \rangle \end{aligned}$$

$\langle \cdot \rangle$  : ensemble average

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# 2 × 2 Hamiltonians



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$\langle \exp[-i\mathbf{C}_1\mathbf{t}] \rangle = \int \exp\left[\frac{(\mathbf{F} - \mathbf{F}^*)}{\mathbf{S}^2}\right] \mathbf{P}(\mathbf{F}, \mathbf{F}^*, \mathbf{S}) d\mathbf{F} d\mathbf{F}^* d\mathbf{S}$$

- matrix elements of  $\mathbf{H}_x, \mathbf{H}_y, \mathbf{H}_0$  correlated in eigenfunction basis of  $\mathbf{H}$

$$\mathbf{0} = \langle \mathbf{1} | \mathbf{H} | \mathbf{2} \rangle = \langle \mathbf{1} | \mathbf{H}_0 | \mathbf{2} \rangle + x \langle \mathbf{1} | \mathbf{H}_x | \mathbf{2} \rangle + y \langle \mathbf{1} | \mathbf{H}_y | \mathbf{2} \rangle$$

→  $\mathbf{F}, \mathbf{F}^*$  are correlated

- is  $\mathbf{S}$  uncorrelated with  $\mathbf{F}, \mathbf{F}^*$ ?

→ eigenvalues and eigenfunctions of  $H$  uncorrelated?

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# H-matrix in site basis: ergodic limit



Statistics of Geometric Phase Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized dynamics

general ensemble

curvature distribution

Details of numerical result

Delocalized Wave Dynamics: waves extended in entire system



In the basis associated with the space

$\langle k | H | l \rangle = H_{kl}$  : Similar Distribution about most probable value

$\langle H_{kl}^2 \rangle$  : Indicator of deterministic uncertainty

If available information :

$$\langle H_{kl}^2 \rangle = \sigma^2, \quad \langle H_{kl} \rangle = 0$$

$$\rho(H) = \exp \left[ -\frac{1}{2\sigma^2} \sum_{k \leq l} H_{kl}^2 \right] = \exp \left[ -\frac{1}{2\sigma^2} \text{Tr} H^2 \right]$$

Wigner-Dyson Ensembles

extended eigenstates:  $\mathbf{H}(x, y) = \mathbf{H}_0 + x \mathbf{H}_x + y \mathbf{H}_y$



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

what is  $P(C)$  when dynamics of the system represented by  $H$  is ergodic?

eigenfunctions delocalized in basis of interest,  $H$ : metallic, chaotic

$H$  : represented by Gaussian unitary ensembles

$$\rho(\mathbf{H}) = \mathcal{N} \exp \left[ -\text{Tr} \mathbf{H}^2 \right]$$

$$\begin{aligned} \mathbf{P}_{e,u}(\mathbf{E}, \mathbf{U}) &= \mathbf{J}(\mathbf{E}, \mathbf{U} | \mathbf{H}) \rho(\mathbf{H}) \\ &= |\mathbf{E}_1 - \mathbf{E}_2|^2 \exp \left[ -(\mathbf{E}_1^2 + \mathbf{E}_2^2) \right] \delta(\mathbf{U}^\dagger \mathbf{U} - \mathbf{I}) \end{aligned}$$

eigenvalues and eigenfunctions of  $H$  not correlated

→ lack of correlations between  $S = |\mathbf{E}_1 - \mathbf{E}_2|$  and  $F$

$$\mathbf{F} = \mathbf{H}_{x;12} \mathbf{H}_{y;21} = \sum_{k,l} \mathbf{U}_{k1}^* \mathbf{U}_{l2} \mathbf{U}_{j2}^* \mathbf{U}_{r1} \mathbf{H}_{x;kl} \mathbf{H}_{y;jr}$$

$$\mathbf{P}(\mathbf{F}, \mathbf{F}^*, \mathbf{S}) = \mathbf{P}_f(\mathbf{F}, \mathbf{F}^*) \mathbf{P}_s(\mathbf{S})$$

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# 2 × 2 Hamiltonians



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

- For mutually independent, Gaussian distributed (zero mean, unit variance) matrix elements of  $H_0, H_x, H_y$  in an arbitrary basis, distribution of  $F \rightarrow$  correlated Gaussian in eigenfunction basis

$$\mathbf{P}_f(\mathbf{F}_x, \mathbf{F}_y) = \mathcal{N} \exp \left[ -\frac{1}{2(1-K^2)} \left( \frac{\mathbf{F}_x^2}{V_x} - \frac{2K\mathbf{F}_x\mathbf{F}_y}{\sqrt{V_x V_y}} + \frac{\mathbf{F}_y^2}{V_y} \right) \right]$$

- $V_x, V_y$ : variances of  $F_x, F_y$ , variances of real and imaginary parts of  $F \equiv F_{12}$
- normalization  $\mathcal{N} = \frac{1}{2\pi\sqrt{V_x V_y(1-K^2)}}$

$$\begin{aligned} \langle \exp[-i\mathbf{C}_1 t] \rangle &= \int \exp \left[ \frac{(\mathbf{F} - \mathbf{F}^*)}{S^2} \right] \mathbf{P}(\mathbf{F}, \mathbf{F}^*) \mathbf{P}_s(S) d\mathbf{F} d\mathbf{F}^* dS \\ &= \int \frac{S^4}{S^4 + 4t^2 V_x V_y (1-K^2)} \mathbf{P}_s(S) dS \end{aligned}$$

# 2 × 2 Hamiltonians



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

**$N=2$  case**

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$P_c(\mathbf{C}) = \frac{1}{2\pi} \int_{-\infty}^{\infty} dt \exp[i\mathbf{C}t] \langle \exp[-i\mathbf{C}_1 t] \rangle$$

- $H_0, H_x, H_y$ : **GUE**  $\Rightarrow$   $H$  **GUE**
- nearest neighbor spacing distribution for **GUE**

$$P_s(\mathbf{S}) = \frac{32}{\pi^2 \langle \mathbf{S} \rangle^3} \mathbf{S}^2 \exp \left[ -\frac{4\mathbf{S}^2}{\pi \langle \mathbf{S} \rangle^2} \right]$$

$$\Rightarrow P_c(\mathbf{C}) = \frac{3}{4A} \left( 1 + \frac{|\mathbf{C}|}{A} \right)^{-5/2}$$

**Rescaling by**  $A = \frac{8\sqrt{V_x V_y (1-K^2)}}{\pi \langle S \rangle^2} = \frac{1}{4(1+x^2+y^2)^{5/2}} \rightarrow$  **universal form**

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# Case $N = 2$ : theory with numerics



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

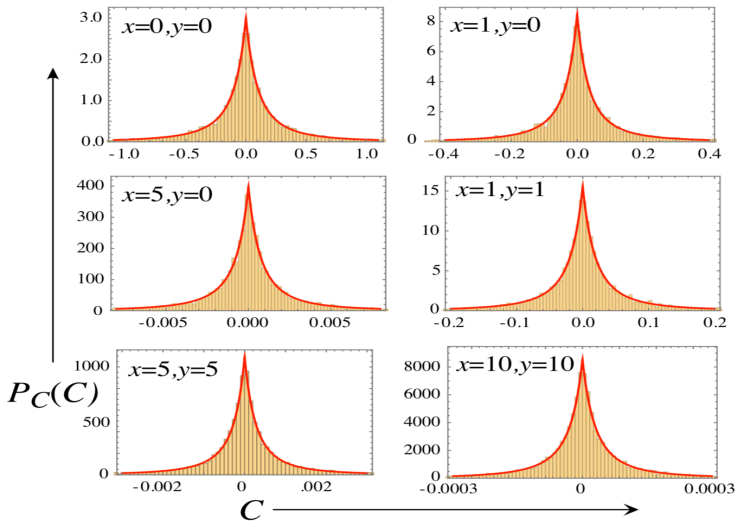
Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result



# $N \times N$ Hamiltonians



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

- $(N - 1)$  states contributing to curvature  $C_n$

$$C_n = 2 \operatorname{Im} \sum_{m \neq n} \frac{\langle n | H_x | m \rangle \langle m | H_y | n \rangle}{(E_n - E_m)^2}$$
$$\approx 2 \operatorname{Im} \left[ \frac{\langle n | H_x | n+1 \rangle \langle n+1 | H_y | n \rangle}{(E_n - E_{n+1})^2} + \frac{\langle n | H_x | n-1 \rangle \langle n-1 | H_y | n \rangle}{(E_n - E_{n-1})^2} \right]$$

- $\langle n | H_x | n+1 \rangle \equiv u_{x+} + i v_{x+}$ ,  $\langle n+1 | H_y | n \rangle \equiv u_{y+} + i v_{y+}$ ,  $E_{n+1} - E_n \equiv S_+$

Desired Probability distribution

$$P_c(C) = \langle \delta [C - C_n] \rangle$$
$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} dt \exp[i C t] \langle \exp[-i C_n t] \rangle$$

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$P_c(c)$ : full form:  $P_C(C) = \frac{1}{A} P_c\left(\frac{C}{A}\right)$



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$P_c(c) = \frac{27\sqrt{3}}{(1+4c^2)^5} \left( \frac{6f_1(|c|)\tan^{-1}(\sqrt{3+8|c|})}{\pi(3+2c^2)^5(3+8|c|)^{9/2}} + \frac{3f_2(|c|)\log(1+2|c|)}{\pi(3+2c^2)^5} \right. \\ \left. + \frac{f_3(|c|)}{2\pi(3+2c^2)^4(3+8|c|)^4} + 3(1-40c^2+80c^4) \right),$$

$$f_1(c) = -33777 - 517914c - 2031912c^2 + 8531280c^3 + \\ 106694400c^4 + 420839296c^5 + 818860032c^6 + 596420096c^7 - \\ 755948800c^8 - 2317539840c^9 - 2585458688c^{10} - 1558491136c^{11} - \\ 472834048c^{12} + 4341760c^{13} + 75366400c^{14} + 46137344c^{15} + \\ 12320768c^{16} + 1572864c^{17},$$

$$f_2(c) = 1 + 2430c + 8120c^2 - 8640c^3 - 57440c^4 - 76224c^5 - \\ 30720c^6 + 15360c^7 + 21760c^8 + 7680c^9 + 2048c^{10},$$

$$f_3(c) = (-41553 - 509328c - 1668384c^2 + 3797712c^3 + 44854416c^4 + \\ 163496512c^5 + 381128320c^6 + 630179840c^7 + 631230720c^8 + \\ 166594560c^9 - 372436992c^{11} - 457797632c^{11} - 216436736c^{12} - \\ 44941312c^{13} + 6979584c^{14} + 1179648c^{14}).$$

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# Case $N = 40$ : theory with numerics



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

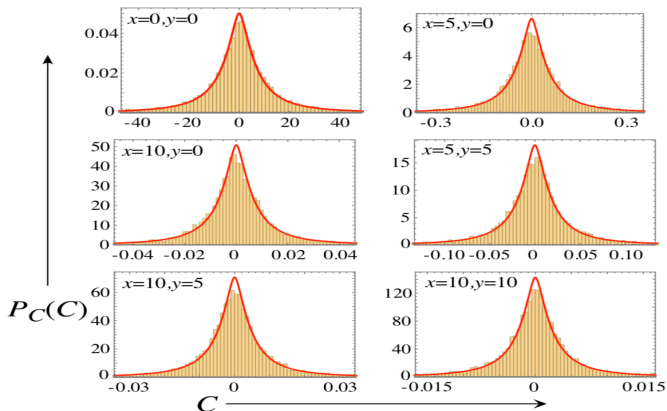
Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result



- slight deviation from theory near  $C = 0$ :  
effect of neglecting non-neighbor levels

# localized/ partially localized eigenstates



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

**what happens when wave-dynamics is not ergodic?**

$H$  : represented by a system-dependent random matrix ensemble

**how the distribution is sensitive to behavior of  $\partial_x \mathbf{H}, \partial_y \mathbf{H}$ ?**

$\partial_x \mathbf{H}, \partial_y \mathbf{H}$ : ensembles different from  $\mathbf{H}$

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# H-matrix in site basis: $\mathbf{H} = \sum_{i,j} \epsilon_i c_i^\dagger \cdot c_i + \sum_{i,j} V_{ij} c_i^\dagger \cdot c_j$



Statistics of Geometric Phase Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

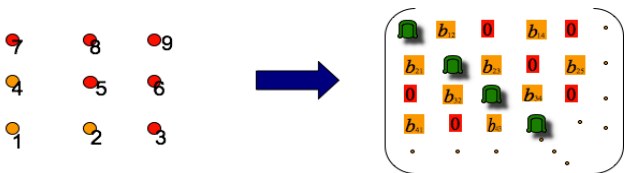
localized dynamics

general ensemble

curvature distribution

Details of numerical result

Localized Wave Dynamics:



In the basis associated with the space

$$1. \quad H_{kl} \neq 0 \quad \text{if} \quad k = l \quad \text{or} \quad k = l \pm L^{d-1}$$

a.  $H_{kk}$  : Randomly Distributed

$$\rho_{kk} = e^{-H_{kk}^2 / 2v_{kk}^2}$$

b. Nearest Neighbor interactions may be random or non random

If non-random  $\rho_{kl} = \delta(H_{kl} - b_{kl})$

If random  $\rho_{kl} = e^{-(H_{kl} - b_{kl})^2 / 2v_{kl}^2}$

Hopping Strength  $b_{kl} = 0$  if  $k \neq l$  or  $k \neq l \pm L^{d-1}$

# system-dependent random matrix ensembles



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

Complex systems represented by multi-parametric Gaussian ensemble

$$\rho(\mathbf{H}, \mathbf{v}, \mathbf{b}) = C \exp \left[ - \sum_{k \leq l} \sum_{s=1}^2 \frac{1}{2v_{kl;s}} (\mathbf{H}_{kl;s} - \mathbf{b}_{kl;s})^2 \right]$$

System information contained in parameter-sets  $\mathbf{v}, \mathbf{b}$ :

$$v_{kl;s}^2(x, y) = \langle \mathbf{H}_{kl;s}(x, y)^2 \rangle - \langle \mathbf{H}_{kl;s}(x, y) \rangle^2, \quad \mathbf{b}_{kl;s}(x, y) = \langle \mathbf{H}_{kl;s}(x, y) \rangle$$

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$$\mathbf{P}_{\mathbf{e}, \mathbf{u}}(\mathbf{E}, \mathbf{U}) = \mathbf{J}(\mathbf{H}|\mathbf{E}, \mathbf{U}) \rho(\mathbf{H}) = \prod_{m \neq n} |\mathbf{E}_m - \mathbf{E}_n|^2 \rho(\mathbf{E}, \mathbf{U})$$

**eigenvalues and eigenfunctions of  $H$  now correlated !**

# curvature distribution for generic case



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$\mathbf{C}_b = 2 \operatorname{Im} \sum_{\mathbf{a} \neq \mathbf{b}} \frac{\langle \mathbf{b} | \partial_x \mathbf{H} | \mathbf{a} \rangle \langle \mathbf{a} | \partial_y \mathbf{H} | \mathbf{b} \rangle}{(\mathbf{E}_a - \mathbf{E}_b)^2} = 2 \operatorname{Im} \sum_{\mathbf{a} \neq \mathbf{b}} \frac{F_{ab}}{(\mathbf{E}_a - \mathbf{E}_b)^2}$$

$$F_{ab} = (\partial_x H)_{ab} \partial_y H_{ba} = \sum_{j,r,k,l} U_{ka}^* U_{lb} U_{jb}^* U_{ra} (\partial_x H)_{kl} \partial_y H_{jr}$$

$$\mathbf{P}_c(\mathbf{C}) = \int \delta(\mathbf{C} - \mathbf{C}_b) \mathbf{P}_{e,u}(\mathbf{U}, \mathbf{U}^*, \mathbf{E}) \rho_x(\partial_x \mathbf{H}) \rho_y(\partial_y \mathbf{H}) D\Gamma$$

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- Integration measure  $D\Gamma \equiv DU DU^* DE D(\partial_x H) D(\partial_y H)$
- $\mathbf{E} \equiv \{E_1, \dots, E_N\}$  and  $F_b \equiv \{F_{1b}, \dots, F_{Nb}\}$  **correlated**

# numerics: three different ensembles



Statistics of Geometric Phase Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

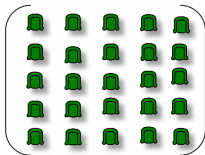
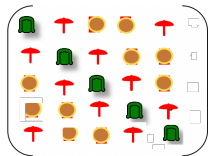
localized dynamics

general ensemble

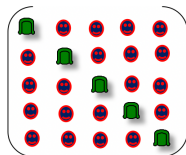
curvature distribution

Details of numerical result

Anderson



GOE/ GUE



Brownian

!0

# perturbation ensemble has any role?



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

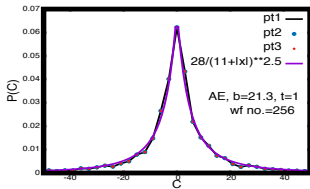
localized  
dynamics

general ensemble

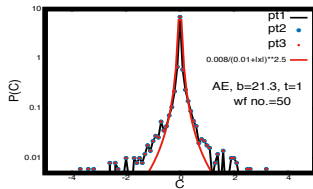
curvature  
distribution

Details of  
numerical result

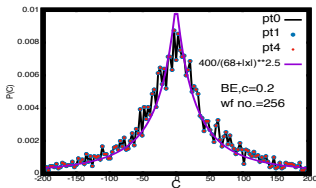
AE, wf=256



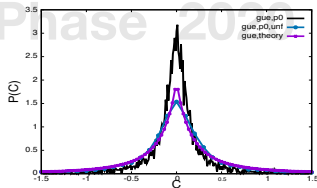
AE, wf=50



BE, wf=256



GUE, wf=256





# comparison: Anderson, Brownian, GUE ensembles



Statistics of Geometric Phase Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

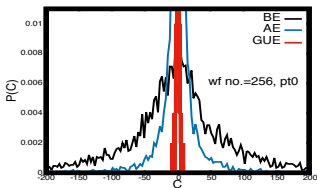
localized dynamics

general ensemble

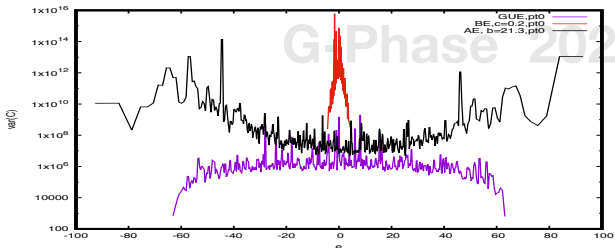
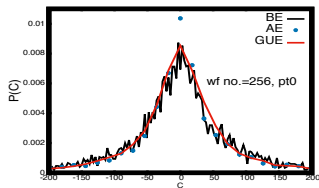
curvature distribution

Details of numerical result

without rescaling



after rescaling



# Single parametric diffusion of $\rho \propto e^{-\sum \frac{(\mathbf{H}_{\mathbf{kl};s} - \mathbf{b}_{\mathbf{kl};s})^2}{2\mathbf{v}_{\mathbf{kl};s}}}$



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

Changing complexity i.e.  $\mathbf{v}$ ,  $\mathbf{b}$  leads to evolution of  $\rho(\mathbf{H})$

possible to define a function  $\mathbf{Y}$  such that (with  $\mathbf{g}_{\mathbf{kl}} = \mathbf{1} + \delta_{\mathbf{kl}}$ )

$$\frac{\partial \rho}{\partial \mathbf{Y}} = \sum_{\mathbf{k}, \mathbf{l}; \mathbf{s}} \frac{\partial}{\partial \mathbf{H}_{\mathbf{kl}; \mathbf{s}}} \left[ \frac{\mathbf{g}_{\mathbf{kl}}}{2} \frac{\partial}{\partial \mathbf{H}_{\mathbf{kl}; \mathbf{s}}} + \mathbf{H}_{\mathbf{kl}; \mathbf{s}} \right] \rho$$

where 
$$\mathbf{Y} = -\frac{1}{\mathbf{N}(\mathbf{N} + 1)} \ln \left[ \prod_{\mathbf{k} \leq \mathbf{l}} \prod_{\mathbf{s}=1}^2 |\mathbf{g}_{\mathbf{kl}} - 2\mathbf{v}_{\mathbf{kl}; \mathbf{s}}| |\mathbf{b}_{\mathbf{kl}; \mathbf{s}}|^2 \right]$$

multi-parametric diffusion of  $\rho(\mathbf{H})$  can be reduced to a single parameter

**$\mathbf{Y}$  = average distribution parameter, a measure of average uncertainty of system**

- moments  $\langle \delta \mathbf{H}_{\mathbf{kl}; \mathbf{s}} \rangle = -\mathbf{H}_{\mathbf{kl}; \mathbf{s}} \delta \mathbf{Y}$ ,  $\langle |\delta \mathbf{H}_{\mathbf{kl}; \mathbf{s}}|^2 \rangle = \mathbf{g}_{\mathbf{kl}} \delta \mathbf{Y}$

# 2nd order perturbation theory



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$\delta \mathbf{e}_n = \delta \mathbf{H}_{nn} + \sum_{m \neq n} \frac{|\delta \mathbf{H}_{mn}|^2}{\mathbf{e}_n - \mathbf{e}_m} + o((\delta \mathbf{H}_{mn})^3)$$

$$\delta U_{jn} = \sum_{m \neq n} \left( \frac{\delta \mathbf{H}_{mn} U_{jm}}{\mathbf{e}_n - \mathbf{e}_m} + \sum_{m'} \frac{\delta \mathbf{H}_{mn} \delta \mathbf{H}_{m'n} U_{jm}}{(\mathbf{e}_n - \mathbf{e}_m)(\mathbf{e}_n - \mathbf{e}_{m'})} - \frac{\delta \mathbf{H}_{mn} \delta \mathbf{H}_{nn}}{(\mathbf{e}_n - \mathbf{e}_m)^2} U_{jm} + \frac{\delta \mathbf{H}_{mn} \delta U_{nm}}{2(\mathbf{e}_n - \mathbf{e}_m)^2} U_{jm} \right)$$

$$\langle \delta \mathbf{e}_n \rangle = \langle \delta \mathbf{H}_{nn} \rangle + \sum_{m \neq n} \frac{\langle |\delta \mathbf{H}_{mn}|^2 \rangle}{\mathbf{e}_n - \mathbf{e}_m} = \left[ \gamma_{\mathbf{e}_n} + \sum_{m \neq n} \frac{1}{\mathbf{e}_n - \mathbf{e}_m} \right] \delta \mathbf{Y}$$

$$\langle \delta \mathbf{e}_n \delta \mathbf{e}_m \rangle = \langle \delta \mathbf{H}_{nn} \delta \mathbf{H}_{mm} \rangle = (2f)^{-1} \delta_{nm} \delta \mathbf{Y}$$

$$\langle \delta U_{jn} \rangle = -\frac{1}{2} \sum_{m=1, m \neq n}^N \frac{U_{jn}}{(\mathbf{e}_n - \mathbf{e}_m)^2} \delta \mathbf{Y}$$

$$\langle \delta U_{jr} \delta U_{kl} \rangle = -\frac{U_{jl} U_{kr}}{(\mathbf{e}_r - \mathbf{e}_l)^2} (1 - \delta_{rl}) \delta \mathbf{Y}$$

$$\langle \delta U_{jr} \delta U_{kl}^* \rangle = \sum_{m \neq n}^N \frac{U_{jm} U_{km}^*}{(\mathbf{e}_r - \mathbf{e}_m)^2} \delta_{rl} \delta \mathbf{Y}$$

# Diffusion Equation for $P_{e,u}(e_1, \dots, e_N; U_1, \dots, U_N; Y)$



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

Using standard second order perturbation theory for Hermitian operators, the second order change due to a small change  $\delta Y$

Joint probability distribution of eigenfunction components  $P_{e,u}$

$$\frac{\partial \mathbf{P}_{e,u}}{\partial \mathbf{Y}} = (\mathbf{L}_e + \mathbf{L}_u + \mathbf{L}_u^*) \mathbf{P}_{e,u}$$

$$\mathbf{L}_e \delta \mathbf{Y} = \sum_n \frac{\partial}{\partial e_n} \left[ \frac{1}{2} \frac{\partial}{\partial e_n} \langle (\delta e_n)^2 \rangle - \langle \delta e_n \rangle \right]$$

$$\mathbf{L}_u \delta \mathbf{Y} = \sum_{k,l,j,r} \frac{\partial}{\partial U_{jr}} \left[ \frac{1}{2} \frac{\partial}{\partial U_{kl}} \langle \delta U_{jr} \delta U_{kl} \rangle + \frac{1}{2} \frac{\partial}{\partial U_{kl}^*} \langle \delta U_{jr} \delta U_{kl}^* \rangle - \langle \delta U_{jr} \rangle \right]$$

# single parametric diffusion of curvature



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$\mathbf{P}_c(\mathbf{C}) = \int \delta(\mathbf{C} - \mathbf{C}_b) \mathbf{P}_{e,u}(\mathbf{U}, \mathbf{U}^*, \mathbf{E}) \rho_x(\partial_x \mathbf{H}) \rho_y(\partial_y \mathbf{H}) D\Gamma$$

$$\frac{\partial \mathbf{P}_c(\mathbf{C})}{\partial \mathbf{Y}} = \int \delta(\mathbf{C} - \mathbf{C}_b) \frac{\partial \mathbf{P}_{e,u}}{\partial \mathbf{Y}} \rho_x(\partial_x \mathbf{H}) \rho_y(\partial_y \mathbf{H}) D\Gamma$$

$$\frac{\partial \mathbf{P}_c(\mathbf{C})}{\partial \mathbf{Y}} = \int \delta(\mathbf{C} - \mathbf{C}_b) (\mathbf{L}_e + \mathbf{L}_u + \mathbf{L}_u^*) \mathbf{P}_{e,u} \rho_x(\partial_x \mathbf{H}) \rho_y(\partial_y \mathbf{H}) D\Gamma$$

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# single parametric diffusion of curvature



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

$$\rho(\mathbf{H}) \propto e^{-\sum \frac{(\mathbf{H}_{\mathbf{k}l;s} - \mathbf{b}_{\mathbf{k}l;s})^2}{2\mathbf{v}_{\mathbf{k}l;s}}}$$

$$\mathbf{Y} = -\frac{1}{\mathbf{N}(\mathbf{N} + 1)} \ln \left[ \prod_{\mathbf{k} \leq l; s} |\mathbf{g}_{\mathbf{k}l} - 2\mathbf{v}_{\mathbf{k}l;s}| |\mathbf{b}_{\mathbf{k}l;s}|^2 \right]$$

For  $2 \times 2$   $H$ -matrix

$$\frac{\partial \mathbf{P}_c(\mathbf{C})}{\partial \mathbf{Y}} = \frac{1}{2} \frac{\partial}{\partial \mathbf{C}} \left[ \alpha \frac{\partial}{\partial \mathbf{C}} + \eta \mathbf{C} \right] \mathbf{P}_c$$

a single parametric formulation of curvature distribution for multi-parametric Gaussian ensemble ( $\alpha, \eta = \text{function}(\text{mean level spacing})$ )

# Story so far



## Statistics of Geometric Phase Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

# G-Phase 2020

# Story so far



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

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## G-Phase 2020



# Story so far



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

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- we analyzed distribution of the geometric curvature for the eigenstates of complex systems represented by a multi-parametric Gaussian ensembles.
- such ensembles can appear whenever system is ergodic/ non-ergodic resulting in a full/ sparse matrix in a physically motivated basis.

## G-Phase 2020

# Story so far



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

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G-Phase 2020

# Story so far



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

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G-Phase 2020

# Story so far



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

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G-Phase 2020

# Story so far



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

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G-Phase 2020

# Story so far



## Statistics of Geometric Phase Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized dynamics

general ensemble

curvature distribution

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- role of fluctuations in systems where phase is topologically invariant?

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# References



Statistics of  
Geometric Phase  
Curvature

Pragya Shukla

Formulation

why needed

Statistics

$N = 2$  case

Motivation

$N=2$  case

$N=2$  case

General  $N$  case

Motivation

localized  
dynamics

general ensemble

curvature  
distribution

Details of  
numerical result

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( $2 \times 2$  case)

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Thank you

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