

# Intrinsic and extrinsic geometries of many body state

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SRH, RS, AC, PRB 98 (23), 235134 (2018)

AC, SRH, RS, PRB,99 (8), 085138 (2019)

SRH, AC, RS, arxiv:1905.13535

- Recap of 1d t-V model.
- Computing distances for t-V model.
- Analysing properties of distances by **Distance Geometry**.
- Computing statistical metric or distance, the **Wasserstein-Kantrovich-Rubinstein** metric.
- Barycenter
- Ollivier-Ricci Curvature.
- Conclusion

- The hamiltonian is

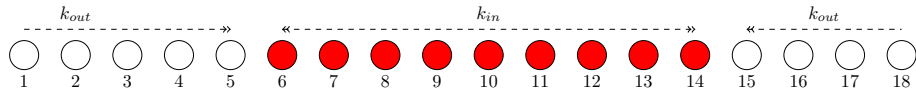
$$H = \sum_i^L (-t(c_i^\dagger c_{i+1} + h.c.) + V n_i n_j)$$

## Recap of phases of 1d t-V model:

- Fermi Liquid at  $V=0$
- Luttinger liquid at  $V$  between  $0^+$  to  $2$
- Luttinger liquid to CDW transition at  $V = V_c = 2$
- CDW for  $V > 2$

# Parameter Space (BZ)

- $c_{k_n} = \sum_i c_i e^{-ik_n i}$  and  $k_n = \frac{2\pi n}{L}$



- $\mathbf{k}_{\text{out}}$ : represents unoccupied states.
- $\mathbf{k}_{\text{in}}$ : represents filled states.
- for  $V=0$

$$|GS\rangle = \prod_{k_n \leq k_f} C_{k_n}^\dagger |0\rangle$$

# Quantum distances for interacting fermions

- Many-body state:

$$|\Psi\rangle = \sum_{\{k\}} C_{k_1, k_2, \dots, k_L} |k_1 k_2 \dots k_L\rangle$$

- Distance between two points  $k_1$  and  $k_2$  on the BZ is computed as :

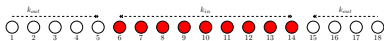
$$d^2(k_1, k_2) = 1 - |\langle \Psi | E(k_1, k_2) | \Psi \rangle|^2$$

where

$$E(k_1, k_2) = e^{\frac{\pi}{2}(c_{k_1}^\dagger c_{k_2} - h.c)}$$

- We can arrange these distances in the form of matrix.

# Distance Matrix as a function of $V$



At  $V = 0$ :

$$D = \left[ \begin{array}{cc|cc} & & k_{in} & k_{out} \\ \hline k_{in} & & 0 & I \\ k_{out} & & I & 0 \end{array} \right]$$

Distance Matrix at  $V = \infty$  (CDW state):

$$D_{ij} = \left[ \begin{array}{cc} 0 & i = j \\ 1 & i = j + \frac{L}{2} \\ \sqrt{\frac{3}{4}} & i \neq j, i \neq j + \frac{L}{2} \end{array} \right]$$

At an arbitrary  $V$ :

$$D = \begin{bmatrix} \Delta & \Delta_e \\ \Delta_e & \Delta \end{bmatrix}$$

## What is Distance Geometry?

**Distance Geometry (DG) is the study of geometry with the basic entity being distance. It all began with the Greeks, specifically Heron, or Hero, of Alexandria sometime between 150BC and 250AD, who showed how to compute the area of a triangle given its side lengths .**

**We human took almost two thousand years to revisit Hero's Formula, finally Arthur Cayley's in 1841 asked the relationships between the distances of five points in space. The gist of what he showed is that a tetrahedron can only exist in a plane if it is flat (in fact, he discussed the situation in one more dimension). This yields algebraic relations on the side lengths of the tetrahedron.**

Karl Menger and Arthur Cayley started this field.

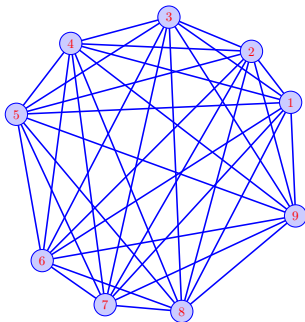
# Geometry: Metric Space (Distance Space)

- We view geometric object as a discrete object.
- This object is defined by set of points and the distances between them.
- We denote all points by set  $V$  and all distances by  $d$ . Together we represent as  $X = (V, d)$
- $X$  is called Metric space or we may call space of distances in a loose sense.



# Geometry : Study of Metric Space or Distance space

- We may view this geometric objects as a graph or network:



- In our case, it is a Ptolemaic metric.
- **The question is what to do with it?**
- **What information one can extract for the physical system?**
- **How to characterize space of distances mathematically?**

## Distance Space: Local view

- We can characterise metric by studying distances and triangles.
- We can define scalar or edge dependent curvatures using distances.
- There are many definition of curvature on a distance space: Menger curvature, Haantjes curvature, Forman Ricci curvature, and Ollivier-Ricci curvature. Many More.

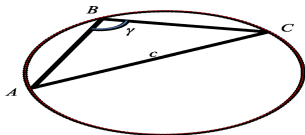
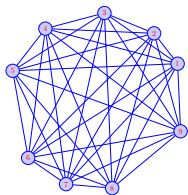


Figure: Menger Curvature at B,  $C(B) = \frac{4A}{d(AB)d(BC)d(AC)}$

# Distance Space: Global structure

- How does a distance space look in Euclidean space ?
- How do we determine the shape of this distance space using only distances? Embedding.
- This can be answered using Distance Geometry. What is distance geometry?
- We can arrange the distances as a matrix  $D$ .

$$\begin{bmatrix} 0 & d_{12} & d_{13} & \dots & d_{1N} \\ d_{21} & 0 & d_{23} & \dots & d_{2N} \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ d_{N1} & d_{N2} & d_{N3} & \dots & d_{NN} \end{bmatrix}$$

## Distance Geometry: Embedding

- We consider Euclidean Space  $R^n$ . Each nodes we represent by coordinate  $x_i, \dots, x_n$ . We ask the following equation:

$$|x_i - x_j|^2 = d_{ij}$$

- Gramian Matrix  $G_{ij} = x_i \cdot x_j$

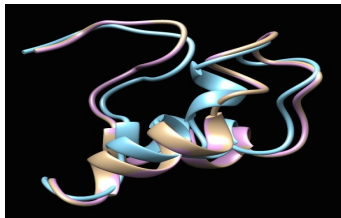
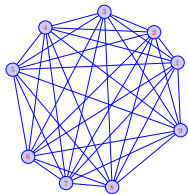
$$G = X^T X$$

- We assume  $x_i$  are mean centered ( $\sum_i x_i = 0$ )
- Matrix G can be expressed in terms of D as

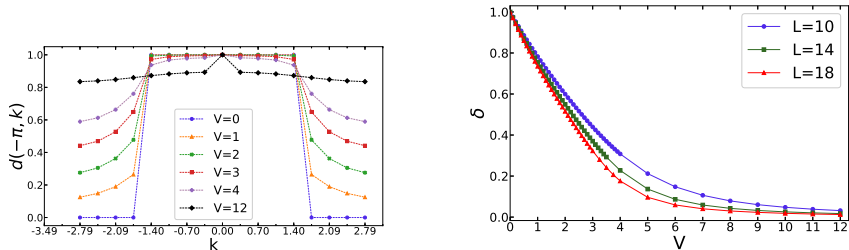
$$G = -\frac{1}{2}JD^2J, J = (I - \frac{ee^T}{n}), e^T = (1, 1, 1, 1 \cdot 1)$$

- If  $D$  can be embedded in the Euclidean space then  $G$  must be positive semi-definite. Its all eigen values must be positive.
- The rank of  $G$  is a dimensionality of the embedding space.
- In most of the cases embedding dimension is very large, so we may embed approximately.
- Non-interacting or Mean field state can be always embedded in finite dimension, but strongly correlated system is embedded in very high dimension.
- Near the phase transition embedding dimension may change.

# Distance Geometry



# Distance Matrix



**Figure:** Distance  $d(-\pi, k)$  between  $k = -\pi$  and the other  $k$  modes in the Brillouin zone (BZ) for different values of the interaction strength  $V$ .

$\delta = d(-\pi, -\pi/2) - d(-\pi, -\pi/2 - 2\pi/L)$ , gives a measure of the discontinuity across the Fermi points. It is studied as a function of interaction strength  $V$  for different system sizes.

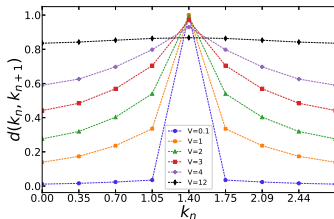


- The Quantum metric  $g(k)$  is defined as:

$$\lim_{\Delta k \rightarrow 0} d^2(k, k + \Delta k) = g(k) \Delta^2 k$$

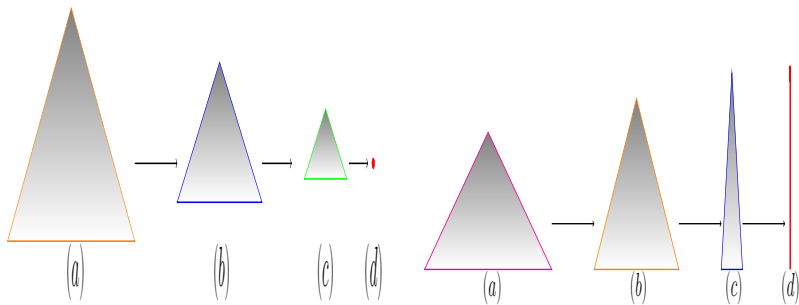
- In this model the distance between two quasi-momentum does not decrease monotonically with the separation.
- The  $V = \infty$  all distances are the same.
- At even small  $V$  there is an optimal value of distance irrespective of the system size. The metric  $g(k)$  is not well defined in this model.

# Nearest Neighbour distance



- $V=0$ : all zero except fermi point (delta function singularity at  $k_f$ )
- intermediate  $V$ : this singularity remains but smoothen out.
- $V = \infty$ : all NN distances are equal.

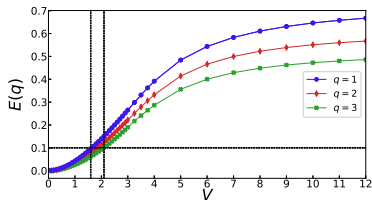
# Behaviour of triangles



**Particle triangles:** It starts shrinking and shrink to points at  $V = 0$ :

**Particle-Hole triangles:** It changes shape at  $V \sim 2$  and become isosceles triangle, they then shrink to segment at  $V = 0$

# Embedding D



**Figure:** Truncation error for  $E(q) = 1.0 - \sqrt{\frac{\sum_{i=0}^q \lambda_i}{\sum_i \lambda_i}}$  keeping first few (1-3) eigenvalues for embedding  $D$ . The truncation error for approximate embedding is less than 12% in case of embedding in one dimension  $E(q=1)$  up to  $V \approx 1.5$  and for embedding in three dimension  $E(q=3)$  up to  $V \approx 2$

# Statistical Distance from Quantum Distances

The statistical distance between any two nodes is defined with respect to  $\Pi_{ij}$  as

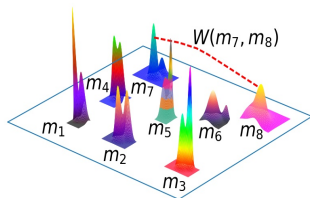
$$W(i, j) = \sum_{l, m} d(l, m) \Pi_{i, j}(l, m)$$

This distance is known as **Wasserstein-Kantrovich-Rubinstein** distance or only **Wasserstien** among Mathematicians.

# Distance distribution on Metric Space (Distance Space)

We define the distance distribution at point  $i$  in the distance space  $M$  as a random walk on  $M$ :

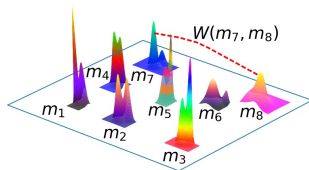
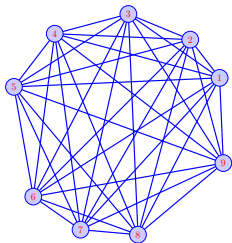
$$m_i(j) = \frac{d_{ij}}{\sum_j d_{ij}}$$



# Probability distribution space

Distance space to

Probability space of Distance distributions:



# Kantorovich Formulation

$$W(m_i, m_j) = \text{Inf}_{\{\pi \in \Pi\}} \sum_{i,j} d(i, j) \pi(i, j)$$

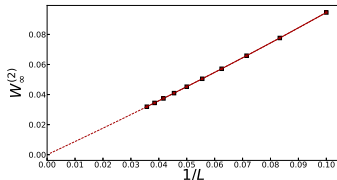
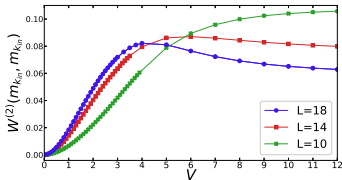
$$\sum_j \pi(i, j) = m_i$$

$$\sum_i \pi(i, j) = m_j$$

- $W(m_i, m_j)$  is distance between distributions  $m_i$  and  $m_j$ .
- $W(m_i, m_j)$  is also known as Earth's Mover Distance (in computer science) and Kantorovich-Wasserstein distance (or metric) among mathematicians. However, Wasserstein has not invented this distance. But it is commonly known Wasserstein distance. Strange world!

**Fields Medalist, Cedric Villani's book:** Optimal Transport, old and new, Springer, 2003



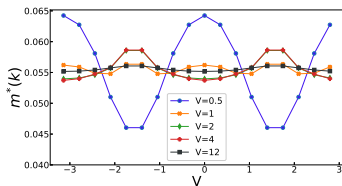


- We can expect  $W$  to indicate the critical interaction strength for the LL to CDW transition, by occurrence of a peak in the Thermodynamic limit.
- We found  $W_{\infty}$  vanishes in the thermodynamic limit. It vanishes as  $1/L$ .

# Barycenter

- In Euclidean case, for a collection of points  $(x_1 \dots x_p)$ , the barycenter  $x^*$  is obtained by minimising the function  $\sum_i^p \lambda_i |x - x_i|^2$ ,  $\sum_i \lambda_i = 1$
- In the present case, the barycenter  $m^*(k)$  is defined as a single function on the BZ as:

$$J(m^*) = \inf_m \frac{1}{L} \sum_i^L W^2(m_i, m)$$

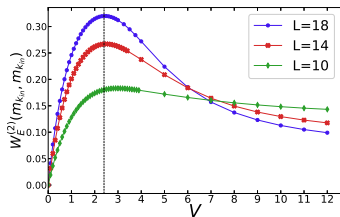


# Quadratic Cost

$$W(m_i, m_j) = \text{Inf}_{\{\pi \in \Pi\}} \sum_{i,j} |i - j|^2 \pi(i, j)$$

$$\sum_j \pi(i, j) = m_i$$

$$\sum_i \pi(i, j) = m_j$$



# Ollivier-Ricci curvature on a distance space

We compute distance between two distributions at  $i$  and  $j$  as:

$$W(m_i, m_j) = \inf_{\pi} \sum_{l,m} d_{lm} \pi_{ij}(l, m)$$

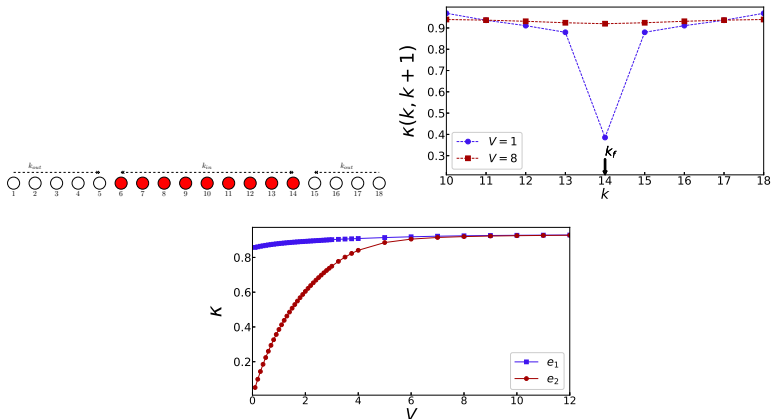
$$\sum_l \pi_{ij}(l, m) = m_j(m), \sum_j \pi_{ij}(l, m) = m_i(l)$$

Ricci-curvature  $K(i, j)$  is defines as:

$$K(i, j) = 1.0 - \frac{W(i, j)}{d_{ij}}$$

Y. Ollivier, C. R. Acad. Sci. Paris, Ser. I 345 (2007).  
(Comptes rendus de l'Academie des Sciences)

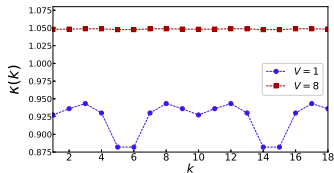
# Ricci-curvature for the nearest neighbour edges



**Figure:** Curvatures for the nearest neighbour edges  $(k, k + 1)$  over half the  $BZ$  for different interaction strengths. The metallic regime is characterised by a discontinuity at the Fermi point  $k_f$ .

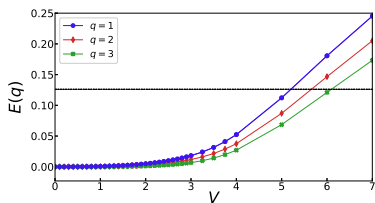
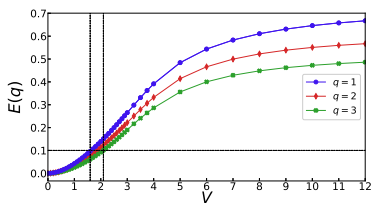
Curvatures for both type of edges  $e_1$  and  $e_2$  as function of interaction strength  $V$ .

# Scalar curvature



**Figure:** Scalar Curvature as a function of the quasi-momenta modes representing vertices of the graph. In insulating regime the scalar curvature is uniform over all the vertices.

# Embedding D and W



**Figure:** Truncation error for keeping first few (1-3) eigenvalues for embedding  $D$ . The truncation error for approximate embedding is less than 12% in case of embedding in one dimension  $E(q=1)$  up to  $V \approx 1.5$  and for embedding in three dimension  $E(q=3)$  up to  $V \approx 2$ . Truncation error for keeping first few (1 – 3) eigenvalues of  $G$  as a function of the interaction strength, for approximate embedding of  $W$ .

# Conclusion

- We have studied properties of distances and tried to extract from it possible cross over transition.
- We also showed using quantum distance and the statistical distance we can probe the shape of the ground state wave function.
- Our expression of the Quantum distance is very general. *can be applied even in one band Hubbard model, which is not possible to study geometry by any other approach such as the Greens function method.*
- *This can be applied to spin system. Work is in progress for MG Model.*
- *Topology can be studied by computing betti numbers of graph of a state or using concept of **magnitude function** of metric space recently developed in enriched category theory. Work in progress.*



**Thank you very much.**