Selberg integrals in 1D random Euclidean optimization problems

Sergio Caracciolo

E-mail: sergio.caracciolo@mi.infn.it Dipartimento di Fisica, University of Milan and INFN, via Celoria 16, 20133 Milan, Italy

Andrea Di Gioacchino

E-mail: andrea.digioacchino@unimi.it Dipartimento di Fisica, University of Milan and INFN, via Celoria 16, 20133 Milan, Italy

Enrico M. Malatesta

E-mail: enrico.m.malatesta@gmail.com Dipartimento di Fisica, University of Milan and INFN, via Celoria 16, 20133 Milan, Italy

Luca G. Molinari

E-mail: luca.molinari@mi.infn.it Dipartimento di Fisica, University of Milan and INFN, via Celoria 16, 20133 Milan, Italy

Abstract. We consider a set of Euclidean optimization problems in one dimension, where the cost function associated to the couple of points x and y is the Euclidean distance between them to an arbitrary power $p \ge 1$, and the points are chosen at random with uniform measure. We derive the exact average cost for the random assignment problem, for any number of points, by using Selberg's integrals. Some variants of these integrals allow to derive also the exact average cost for the bipartite travelling salesman problem.

1. Selberg integrals

Euler Beta integrals

Beta
$$(\alpha, \beta) := \int_{0}^{1} dx \, x^{\alpha-1} (1-x)^{\beta-1} = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$$
 (1.1)

with $\alpha, \beta \in \mathbb{C}$ and $\Re(\alpha) > 0$, $\Re(\beta) > 0$, have been generalized in the 1940s by Atle Selberg [1]

$$S_n(\alpha,\beta,\gamma) := \left(\prod_{i=1}^n \int_0^1 dx_i \, x_i^{\alpha-1} (1-x_i)^{\beta-1}\right) |\Delta(x)|^{2\gamma}$$

$$= \prod_{j=1}^n \frac{\Gamma(\alpha+(j-1)\gamma)\Gamma(\beta+(j-1)\gamma)\Gamma(1+j\gamma)}{\Gamma(\alpha+\beta+(n+j-2)\gamma)\Gamma(1+\gamma)}$$
(1.2)

where

$$\Delta(x) := \prod_{1 \le i < j \le n} (x_i - x_j) \tag{1.3}$$

with $\alpha, \beta, \gamma \in \mathbb{C}$ and $\Re(\alpha) > 0$, $\Re(\beta) > 0$, $\Re(\gamma) > \min(1/n, \Re(\alpha)/(n-1), \Re(\beta)/(n-1))$, see [2, Chap. 8]. Indeed (1.2) reduces to (1.1) when n = 1.

These integrals have been used by Enrico Bombieri (see [3] for the detailed history) to prove what was known as the *Mehta-Dyson conjecture* [4–7] in random matrix theory, that is

$$F_n(\gamma) := \left(\prod_{i=1}^n \int_{-\infty}^{+\infty} \frac{dx_i}{\sqrt{2\pi}} e^{-\frac{x_i^2}{2}}\right) |\Delta(x)|^{2\gamma} = \prod_{j=1}^n \frac{\Gamma(1+j\gamma)}{\Gamma(1+\gamma)}$$
(1.4)

but they found many applications also in the context of conformal field theories [8] and exactly solvable models, for example in the evaluation of the norm of *Jack polynomials* [9]. In the following we shall also need of an extension of Selberg integrals [2, Sec. 8.3]

$$B_{n}(j,k;\alpha,\beta,\gamma) := \left(\prod_{i=1}^{n} \int_{0}^{1} dx_{i} x_{i}^{\alpha-1} (1-x_{i})^{\beta-1}\right) \left(\prod_{s=1}^{j} x_{s}\right) \left(\prod_{s=j+1}^{j+k} (1-x_{s})\right) |\Delta(x)|^{2\gamma}$$
$$= S_{n}(\alpha,\beta,\gamma) \frac{\prod_{i=1}^{j} [\alpha+(n-i)\gamma] \prod_{i=1}^{k} [\beta+(n-i)\gamma]}{\prod_{i=1}^{j+k} [\alpha+\beta+(2n-1-i)\gamma]}.$$
(1.5)

In this paper we present one more application of Selberg integrals in the context of Euclidean random combinatorial optimization problems in one dimension. While these problems have been well understood in the case in which the entries of the cost matrix, that is the costs between couple of points, are independent and equally distributed random variables, also thanks to the methods developed in the context of statistical mechanics of disordered systems [10], the interesting physical case in which the cost matrix is a Euclidean random matrix [11] is much less studied because of the difficulties induced by correlations among the entries. The possibility of dealing with correlations as a perturbation has been investigated [12] and is effective for very large dimension of the Euclidean space. On the opposite side, detailed results, at least in the limit of an asymptotic number of points, have been recently obtained in one dimension [13–17,19] and in two dimensions [18,20–23]. Note that related problems in higher than one dimension have also been studied from a rigorous point of view [24–26].

We shall concentrate here in the one dimensional case, with the aim at proving new exact results, which we have obtained by exploiting the Selberg integrals. Let us

$$P_k(x)dx := \frac{\Gamma(N+1)}{\Gamma(k)\,\Gamma(N-k+1)}\,x^{k-1}(1-x)^{N-k}dx\,.$$
(1.6)

To every couple x_i and x_j we associate a cost which depends only on their Euclidean distance, in the form

$$w_{ij}^{(p)} = f(|x_i - x_j|) := |x_i - x_j|^p$$
(1.7)

for some p > 1, so that the function f is an increasing and convex function of its argument.

The new results presented in this paper regard finite-size systems. The paper is organized as follows. In Sec. 2, respectively Sec. 3, we present the evaluation of the average cost for the assignment, respectively the TSP problem, for all values of the parameter p > 1 and all numbers of points. In Sec. 4 we evaluate, instead, under the same conditions, what is lost in the average cost if we do not match the k-th point in the best way, for each k. This is needed if we want to compute upper bounds to the average cost of the 2-factor problem.

2. Average cost in the assignment

Consider two sets of N points chosen at random in the interval [0, 1], the red $\{r_i\}_{\in [N]}$, and the blue $\{b_i\}_{i\in [N]}$, points. In the assignment problem we have to find the one-toone correspondence between the red and blue points, that is a permutation σ in the symmetric group of N elements S_N , such that the total cost

$$E_N^{(p)}[\sigma] := \sum_{i=1}^N w_{i,\sigma(i)}^{(p)} = \sum_{i=1}^N |r_i - b_{\sigma(i)}|^p$$
(2.1)

is minimized.

When p > 1 the optimal solution is the identity permutation, once both sets of points have been ordered [13, 27]. It follows that, in these cases, the optimal cost is

$$E_N^{(p)} = \sum_{i=1}^N |r_i - b_i|^p \tag{2.2}$$

By using (1.6) and the Selberg integral (1.2)

$$S_{2}\left(k, N-k+1, \frac{p}{2}\right) = \left(\prod_{i=1}^{2} \int_{0}^{1} dx_{i} x_{i}^{k-1} (1-x_{i})^{N-k}\right) |x_{2}-x_{1}|^{p}$$
$$= \frac{\Gamma(k)\Gamma(N-k+1)\Gamma\left(k+\frac{p}{2}\right)\Gamma\left(N-k+1+\frac{p}{2}\right)\Gamma(1+p)}{\Gamma\left(N+1+\frac{p}{2}\right)\Gamma(N+1+p)\Gamma\left(1+\frac{p}{2}\right)}.$$
(2.3)

we get that the average of the k-th contribution is given by

$$\overline{|r_k - b_k|^p} = \int_0^1 dx \int_0^1 dy P_k(x) P_k(y) |y - x|^p$$

= $\left(\frac{\Gamma(N+1)}{\Gamma(k)\Gamma(N-k+1)}\right)^2 \left(\prod_{i=1}^2 \int_0^1 dx_i x_i^{k-1} (1-x_i)^{N-k}\right) |x_2 - x_1|^p$ (2.4)
= $\left(\frac{\Gamma(N+1)}{\Gamma(k)\Gamma(N-k+1)}\right)^2 S_2 \left(k, N-k+1, \frac{p}{2}\right)$
= $\frac{\Gamma^2(N+1)\Gamma\left(k+\frac{p}{2}\right)\Gamma\left(N-k+1+\frac{p}{2}\right)\Gamma(1+p)}{\Gamma(k)\Gamma(N-k+1)\Gamma\left(N+1+\frac{p}{2}\right)\Gamma(N+1+p)\Gamma\left(1+\frac{p}{2}\right)}$

and therefore we get the exact result

$$\overline{E_N^{(p)}} = \frac{\Gamma^2(N+1)\Gamma(1+p)}{\Gamma\left(N+1+\frac{p}{2}\right)\Gamma(N+1+p)\Gamma\left(1+\frac{p}{2}\right)} \sum_{k=1}^N \frac{\Gamma\left(k+\frac{p}{2}\right)\Gamma\left(N-k+1+\frac{p}{2}\right)}{\Gamma(k)\Gamma(N-k+1)}$$

$$= \frac{\Gamma\left(1+\frac{p}{2}\right)}{p+1} N \frac{\Gamma(N+1)}{\Gamma\left(N+1+\frac{p}{2}\right)}$$
(2.5)

where we made repeated use of the duplication and Euler's inversion formula for $\Gamma\text{-}$ functions

$$\Gamma(z)\Gamma\left(z+\frac{1}{2}\right) = 2^{1-2z}\sqrt{\pi}\,\Gamma(2z) \tag{2.6a}$$

$$\Gamma(1-z)\Gamma(z) = \frac{\pi}{\sin(\pi z)}.$$
(2.6b)

The exact result (2.4) was known only in the cases p = 2, 4 where the computation can be carried out simply by using the Euler Beta function (1.1), see [15]. With $p \neq 2, 4$, the average optimal cost has been computed only in the limit of large N [15] at the order $o(N^{-1})$. From (2.4) we easily get also the next correction in 1/N

$$\overline{E_N^{(p)}} = \frac{\Gamma\left(1+\frac{p}{2}\right)}{p+1} N^{1-\frac{p}{2}} \left(1 - \frac{p(p+2)}{8N} + \frac{p(p+2)(p+4)(3p+2)}{384N^2} + o(N^{-2})\right).$$
(2.7)

3. Average cost in the TSP

Again consider two sets of N points chosen at random in the interval [0, 1], the red $\{r_i\}_{\in[N]}$, and the blue $\{b_i\}_{i\in[N]}$, points. In the Travelling Salesman Problem (TSP) we have to choose a closed path which visits only once all the 2N points, alternating red and blue points along the walk, that is, see [16], two permutations σ and π in the symmetric group of N elements S_N , such that the total cost

$$E_N^{(p)}[\sigma] := \sum_{i=1}^N [w_{\sigma(i),\pi(i)}^{(p)} + w_{\sigma\circ\tau(i),\pi(i)}^{(p)}] = \sum_{i=1}^N \left[|r_{\sigma(i)} - b_{\pi(i)}|^p + |r_{\sigma\circ\tau(i)} - b_{\pi(i)}|^p \right]$$
(3.1)

is minimal. In the previous equation τ is the shifting permutation, that is $\tau(i) = i+1$ for $i \in [N-1]$ and $\tau(N) = 1$.

When p > 1 the optimal solution is given by the permutations [16]

$$\tilde{\sigma}(i) = \begin{cases} 2i - 1 & i \le (N+1)/2\\ 2N - 2i + 2 & i > (N+1)/2 \end{cases}$$
(3.2)

 $\quad \text{and} \quad$

$$\tilde{\pi}(i) = \tilde{\sigma}(N+1-i) = \begin{cases} 2i & i < (N+1)/2\\ 2N-2i+1 & i \ge (N+1)/2 \end{cases}$$
(3.3)

once both sets of points have been ordered. It follows that, in these cases, the optimal cost is \$N-1\$

$$E_N^{(p)} = |r_1 - b_1|^p + |r_N - b_N|^p + \sum_{i=1}^{N-1} [|r_{i+1} - b_i|^p + |r_i - b_{i+1}|^p]$$
(3.4)

By using (1.6) and the generalized Selberg integral (1.5)

$$B_{2}\left(1,1;k,N-k,\frac{p}{2}\right) = \\ = \int_{0}^{1} dx_{1} \int_{0}^{1} dx_{2} x_{1}^{k-1} x_{2}^{k} (1-x_{1})^{N-k} (1-x_{2})^{N-k-1} |x_{1}-x_{2}|^{p} \\ = \frac{\left(k+\frac{p}{2}\right) \left(N-k+\frac{p}{2}\right)}{\left(N+p\right) \left(N+\frac{p}{2}\right)} S_{2}\left(k,N-k,\frac{p}{2}\right) \\ = \frac{\Gamma(k)\Gamma(N-k)\Gamma(p+1)\Gamma\left(k+\frac{p}{2}+1\right)\Gamma\left(N-k+\frac{p}{2}+1\right)}{\Gamma(N+p+1)\Gamma\left(N+\frac{p}{2}+1\right)\Gamma\left(1+\frac{p}{2}\right)}$$
(3.5)

we get

$$\begin{aligned} \overline{|b_{k+1} - r_k|^p} &= \overline{|r_{k+1} - b_k|^p} = \int_0^1 dx \int_0^1 dy \, P_k(x) \, P_{k+1}(y) \, |x - y|^p \\ &= \frac{\Gamma^2(N+1)}{\Gamma(k) \, \Gamma(N-k) \, \Gamma(k+1) \, \Gamma(N-k+1)} \times \\ &\times \int_0^1 dx \, dy \, x^{k-1} \, y^k (1 - x)^{N-k} (1 - y)^{N-k-1} \, |x - y|^p \end{aligned}$$
(3.6)
$$&= \frac{\Gamma^2(N+1)}{\Gamma(k) \, \Gamma(N-k) \, \Gamma(k+1) \, \Gamma(N-k+1)} \, B_2\left(1, 1; k, N-k, \frac{p}{2}\right) \\ &= \frac{\Gamma^2(N+1) \, \Gamma(p+1) \, \Gamma\left(k + \frac{p}{2} + 1\right) \, \Gamma\left(N-k + \frac{p}{2} + 1\right)}{\Gamma(k+1) \, \Gamma(N-k+1) \, \Gamma(N+p+1) \, \Gamma\left(N + \frac{p}{2} + 1\right) \, \Gamma\left(1 + \frac{p}{2}\right)}. \end{aligned}$$

from which we obtain

$$\sum_{k=1}^{N-1} \overline{|b_{k+1} - r_k|^p} = 2\Gamma(N+1)\Gamma(1+p) \left(\frac{(N+p+1)\Gamma(\frac{p}{2})}{4(p+1)\Gamma(p)\Gamma(N+1+\frac{p}{2})} - \frac{1}{\Gamma(N+1+p)}\right).$$
(3.7)



Figure 1. Graphical representation of the cutting operation which brings from the optimal TSP cycle (top) to a possible optimal solution of the 2-factor problem (bottom). Here we have represented the N = 4 case, where the cutting operation is unique. Notice that blue and red points are chosen on a interval, but here they are represented equispaced on two parallel lines to improve visualization.

In addition

$$\overline{|r_1 - b_1|^p} = \overline{|r_N - b_N|^p} = N^2 \int_0^1 dx \, dy \, (xy)^{N-1} \, |x - y|^p$$

$$= N^2 S_2 \left(N, 1, \frac{p}{2}\right) = \frac{N \, \Gamma(N+1) \, \Gamma(p+1)}{\left(N + \frac{p}{2}\right) \, \Gamma(N+p+1)} \,.$$
(3.8)

Finally, the average optimal cost for every N and every p > 1 is

$$\overline{E_N^{(p)}} = 2\left(\overline{|r_1 - b_1|^p} + \sum_{k=1}^{N-1} \overline{|b_{k+1} - r_k|^p}\right) = 2\Gamma(N+1)\left[\frac{(N+p+1)\Gamma(1+\frac{p}{2})}{(p+1)\Gamma(N+1+\frac{p}{2})} - \frac{2\Gamma(p+1)}{(2N+p)\Gamma(N+p)}\right].$$
(3.9)

For p = 2 this reduces to

$$\overline{E}_N^{(2)} = \frac{2}{3} \frac{N^2 + 4N - 3}{(N+1)^2} \,. \tag{3.10}$$

which was already known [16]. For generic p > 1 only the asymptotic behaviour for large N was obtained in [16]

$$\lim_{N \to \infty} N^{p/2 - 1} \overline{E_N^{(p)}} = 2 \, \frac{\Gamma\left(\frac{p}{2} + 1\right)}{p + 1} \,, \tag{3.11}$$

which agrees with the large N limit of Eq. (3.9).

4. Cutting shoelaces: the 2-factor problem

In the 2-factor, or 2-assignment problem we carry out the minimization on the set of all the spanning subgraphs in which each vertex is shared by two edges. In different



Figure 2. Plot of $E_k^{(p)}$ given in Eq. (4.4) for various values of p: the green line is calculated with p = 2.1, the orange with p = 2 and the blue one with p = 1.9; in all cases we take N = 100.

words, this problem corresponds to a loop covering of the graph, i.e. we relax the unique cycle condition we had in the TSP. As shown in [17], for every value of N, the optimal 2-factor solution is always composed by an union of shoelaces loops with only two or three points of each color. As a consequence, differently from the assignment and the TSP cases, different instances of the disorder can have different spanning subgraphs that minimize the cost function. In particular these spanning subgraphs can always be obtained by "cutting" the optimal TSP cycle (see Fig. 1) in a way which depends on the specific instance. This "instance dependence" makes the computation of the average optimal cost particularly difficult. However, one can show that the average optimal cost of the 2-factor problem is bounded from above by the TSP average optimal cost and from below by twice the assignment average optimal cost. Since in the large N limit these two quantities coincide, one obtains immediately the large N limit of the average optimal cost of the 2-factor problem. Unfortunately, this approach is not useful for a finite-size system. But we can use Selberg integrals to obtain an upper bound: indeed we can compute the average cost obtained by "cutting" the TSP optimal cycle in specific ways. When we cut at the k-position the optimal TSP into two different cycles we gain an average cost

$$E_{k}^{(p)} = \overline{|b_{k+1} - r_{k}|^{p}} + \overline{|r_{k+1} - b_{k}|^{p}} - \overline{|b_{k+1} - r_{k+1}|^{p}} - \overline{|b_{k} - r_{k}|^{p}}.$$
 (4.1)

Once more, by using (1.6) and the generalized Selberg integral (1.5), we obtain

$$\overline{|b_k - r_k|^p} - \overline{|b_{k+1} - r_k|^p} =
= \frac{\Gamma^2(N+1)\Gamma(p+1)\Gamma(k+\frac{p}{2})\Gamma(N-k+\frac{p}{2}+1)}{\Gamma(k)\Gamma(N-k+1)\Gamma(N+p+1)\Gamma(N+\frac{p}{2}+1)\Gamma(\frac{p}{2}+1)} \left[1 - \frac{k+\frac{p}{2}}{k}\right] \quad (4.2)
= -\frac{p}{2}\frac{\Gamma^2(N+1)\Gamma(p+1)\Gamma(k+\frac{p}{2})\Gamma(N-k+\frac{p}{2}+1)}{\Gamma(k+1)\Gamma(N-k+1)\Gamma(N+p+1)\Gamma(N+\frac{p}{2}+1)\Gamma(\frac{p}{2}+1)},$$

and similarly

$$\overline{|b_{k+1} - r_{k+1}|^p} - \overline{|r_{k+1} - b_k|^p} = \frac{\Gamma^2(N+1)\Gamma(p+1)\Gamma(k+\frac{p}{2}+1)\Gamma(N-k+\frac{p}{2})}{\Gamma(k+1)\Gamma(N-k)\Gamma(N+p+1)\Gamma(N+\frac{p}{2}+1)\Gamma(\frac{p}{2}+1)} \left[1 - \frac{N-k+\frac{p}{2}}{N-k}\right] (4.3)$$
$$= -\frac{p}{2}\frac{\Gamma^2(N+1)\Gamma(p+1)\Gamma(k+\frac{p}{2}+1)\Gamma(N-k+\frac{p}{2})}{\Gamma(k+1)\Gamma(N-k+1)\Gamma(N+p+1)\Gamma(N+\frac{p}{2}+1)\Gamma(\frac{p}{2}+1)}.$$

Their sum is

$$E_{k}^{(p)} = \frac{p}{2} \frac{\Gamma^{2}(N+1)\Gamma(p+1)\Gamma\left(k+\frac{p}{2}\right)\Gamma\left(N-k+\frac{p}{2}\right)}{\Gamma(k+1)\Gamma(N-k+1)\Gamma(N+p)\Gamma\left(N+\frac{p}{2}+1\right)\Gamma\left(\frac{p}{2}+1\right)},$$
(4.4)

For p = 2 this quantity is in agreement with what we got in [17]

$$E_k^{(2)} = \frac{2}{(N+1)^2}.$$
(4.5)

For $p \neq 2$, E_k depends on k. In particular, for 1 the cut near to 0 and 1 are (on average) more convenient than those near the center. For <math>p > 2 the reverse is true (see Fig. 2). Notice that for p = 2 one can see that the best upper bound for the average optimal cost is given by summing the maximum number of cuts that can be done on the optimal TSP cycle. For $p \neq 2$, however, this sum does not give a simple formula.

5. Conclusions

In this work we have been able to obtain some finite-size properties of a set of bipartite Euclidean optimization problems: the assignment, the bipartite TSP and the bipartite 2-factor problems by extensive use of the Selberg integrals. This confirms once more their importance and their wide application range. Interestingly, we have confirmed that the one dimensional 2-factor problem is more subtle to deal with than the other problems considered here, because even using Selberg integrals, we have not been able to find a finite-size upper-bound for the average optimal cost in the generic $p \neq 2$ case. However, our approach allowed to understand the source of this difficulty: for 1 , the shortest loops tend to concentrate at the border of the [0, 1] interval, while for <math>p > 2 they tend to concentrate in the center of it.

References

- [1] Atle Selberg. Remarks on a multiple integral. Norsk Mat. Tidsskr., 26(71-78), 1944.
- [2] George E. Andrews, Richard Askey, and Ranjan Roy. Special Functions, volume 71 of Encyclopedia of Mathematics and its Applications. Cambridge University Press, 1999.
- [3] Peter J. Forrester and S. Ole Warnaar. The importance of the Selberg integral. Amer. Math. Soc. (N. S.), 45(4):489–534, 2008.
- M. L. Mehta and Freeman J. Dyson. Statistical theory of the energy levels of complex systems.
 v. J. Math. Phys., 4:713-719, 1963.
- [5] M. L. Mehta. Random Matrices and the Statistical Theory of Energy Levels. AcademicPress, New York, 1967.
- [6] M. L. Mehta. Problem 746, three multiple integrals. SIAM Review, 16:256–257, 1974.
- [7] M. L. Mehta. Random Matrices, volume 142 of Pure and Applied Mathematics. Elsevier/Academic Press, Amsterdam, 2004.

- [8] V. S. Dotsenko and V. A. Fateev. Four-point correlation functions and the operator algebra in 2d conformal invariant theories with central charge $c \leq 1$. Nucl. Phys. B, 251:691–734, 1985.
- [9] S. Kakei. Intertwining operators for a degenerate double affine Hecke algebra and multivariable orthogonal polynomials. J. Math. Phys., 39:4993–5006, 1998.
- [10] Marc Mézard, Giorgio Parisi, and Miguel Virasoro. Spin glass theory and beyond: An Introduction to the Replica Method and Its Applications, volume 9. World Scientific Publishing Company, 1987.
- [11] Marc Mézard, Giorgio Parisi, and Anthony Zee. Spectra of euclidean random matrices. Nucl. Phys. B, 559(3):689–701, 1999.
- [12] Marc Mézard and Giorgio Parisi. The euclidean matching problem. Journal de Physique, 49:2019–2025, 1988.
- [13] Elena Boniolo, Sergio Caracciolo, and Andrea Sportiello. Correlation function for the gridpoisson euclidean matching on a line and on a circle. J. Stat. Mech., 11:P11023, 2014.
- [14] Sergio Caracciolo and Gabriele Sicuro. On the one dimensional euclidean matching problem: exact solutions, correlation functions and universality. *Phys. Rev. E*, 90(042112), 2014.
- [15] Sergio Caracciolo, Matteo P. D'Achille, and Gabriele Sicuro. Random euclidean matching problems in one dimension. *Phys. Rev. E*, 96:042102, 2017.
- [16] Sergio Caracciolo, Andrea Di Gioacchino, Marco Gherardi, and Enrico M. Malatesta. Solution for a bipartite euclidean tsp in one dimension. *Phys. Rev. E*, 97(5):052109, 2018.
- [17] Sergio Caracciolo, Andrea Di Gioacchino, and Enrico M. Malatesta. Plastic number and possible optimal solutions for an euclidean 2-matching in one dimension. J. Stat. Mech., page 083402, 2018.
- [18] Sergio Caracciolo, Gabriele Sicuro. Quadratic Stochastic Euclidean Bipartite Matching Problem. Phys. Rev. Lett., 115:230601, 2015.
- [19] Sergio Caracciolo, Andrea Di Gioacchino, Enrico M. Malatesta, and Carlo Vanoni. Monopartite euclidean tsp in one dimension. accepted in J. Phys. A: Math. Theor., 2019.
- [20] Sergio Caracciolo, Carlo Lucibello, Giorgio Parisi, and Gabriele Sicuro. Scaling hypothesis for the euclidean bipartite matching problem. *Phys. Rev. E*, 90:012118, 2014.
- [21] Sergio Caracciolo and Gabriele Sicuro. Scaling hypothesis for the euclidean bipartite matching problem. ii. correlation functions. *Phys. Rev. E*, 91(062125), 2015.
- [22] Riccardo Capelli, Sergio Caracciolo, Andrea Di Gioacchino, and Enrico M. Malatesta. Exact value for the average optimal cost of bipartite traveling-salesman and 2-factor problems in two dimensions. *Phys. Rev. E*, 2018.
- [23] Luigi Ambrosio, Federico Stra, and Dario Trevisan. A PDE approach to a 2-dimensional matching problem. Probab. Theory Relat. Fields, 173:433, 2018.
- [24] M Talagrand. The transportation cost from the uniform measure to the empirical measure in dimension 3. The Annals of Probability, 22, pages 919-959, 1994.
- [25] Nicolás Garcia Trillos and Dejan Slepčev. On the rate of convergence of empirical measures in-transportation distance. *Canadian Journal of Mathematics*, 67(6):1358–1383, 2015.
- [26] Nicolás García Trillos and Dejan Slepčev. Continuum limit of total variation on point clouds. Archive for Rational Mechanics and Analysis, 220(1):193–241, Apr 2016.
- [27] Robert J. McCann. Exact solutions to the transportation problem on the line. Proc. R. Soc. A: Math., Phys. Eng. Sci., 455:1341–1380, 1999.