

Entropic-graphs: Theory

Alfred O. Hero

Dept. EECS, Dept Biomed. Eng., Dept. Statistics

University of Michigan - Ann Arbor

`hero@eecs.umich.edu`

`http://www.eecs.umich.edu/~hero`

Collaborators: H. Heemuchwala, J. Costa, B. Ma, O. Michel

- Rényi α -Entropy and Rényi α -Divergence
- Minimal graphs and entropic graphs
- Entropic graphs: asymptotic convergence results
- Extensions to partitioning approximations
- Open problems

Rényi Entropy and Rényi Divergence

- $X \sim f(x)$ a d -dimensional random vector.
- Rényi Entropy of order α

$$H_{\alpha}(f) = \frac{1}{1-\alpha} \ln \int f^{\alpha}(x) dx \quad (1)$$

- The Rényi α -divergence of fractional order $\alpha \in [0, 1]$ [Rényi:61,70]

$$\begin{aligned} D_{\alpha}(f_1 \parallel f_0) &= \frac{1}{\alpha-1} \ln \int f_1 \left(\frac{f_1}{f_0} \right)^{\alpha} dx \\ &= \frac{1}{\alpha-1} \ln \int f_1^{\alpha} f_0^{1-\alpha} dx \end{aligned}$$

- α -Divergence vs α -Entropy

$$H_{\alpha}(f_1) = \frac{1}{1-\alpha} \ln \int f_1^{\alpha} dx = -D_{\alpha}(f_1 \parallel f_0)|_{f_0=U([0,1]^d)}$$

- α -Divergence vs. Batthacharyya-Hellinger distance

$$\begin{aligned} D_{BH}^2(f_1 \parallel f_0) &= \int \left(\sqrt{f_1} - \sqrt{f_0} \right)^2 dx \\ &= 2 \left(1 - \exp \left(\frac{1}{2} D_{\frac{1}{2}}(f_1 \parallel f_0) \right) \right) \end{aligned}$$

- α -Divergence vs. Kullback-Liebler divergence (Shannon MI)

$$\lim_{\alpha \rightarrow 1} D_{\alpha}(f_1 \parallel f_0) = \int f_1 \ln \frac{f_1}{f_0} dx.$$

Entropic Graphs

A graph G of degree l consists of vertices and edges

- vertices are subset of $X_n = \{x_i\}_{i=1}^n$: n points in \mathbf{R}^d
- edges are denoted $\{e_{ij}\}$
- for any i : $\text{card}\{e_{ij}\}_j \leq l$

Weight (with power exponent γ) of G

$$L_\gamma^G(X_n) = \sum_{e \in G} \|e\|^\gamma$$

Minimal Spanning Tree (MST)

Let $T_n = T(X_n)$ denote the possible sets of edges in the class of acyclic graphs spanning X_n (spanning trees).

The Euclidean Power Weighted MST achieves

$$L_\gamma^{\text{MST}}(X_n) = \min_{T_n} \sum_{e \in T_n} \|e\|^\gamma.$$

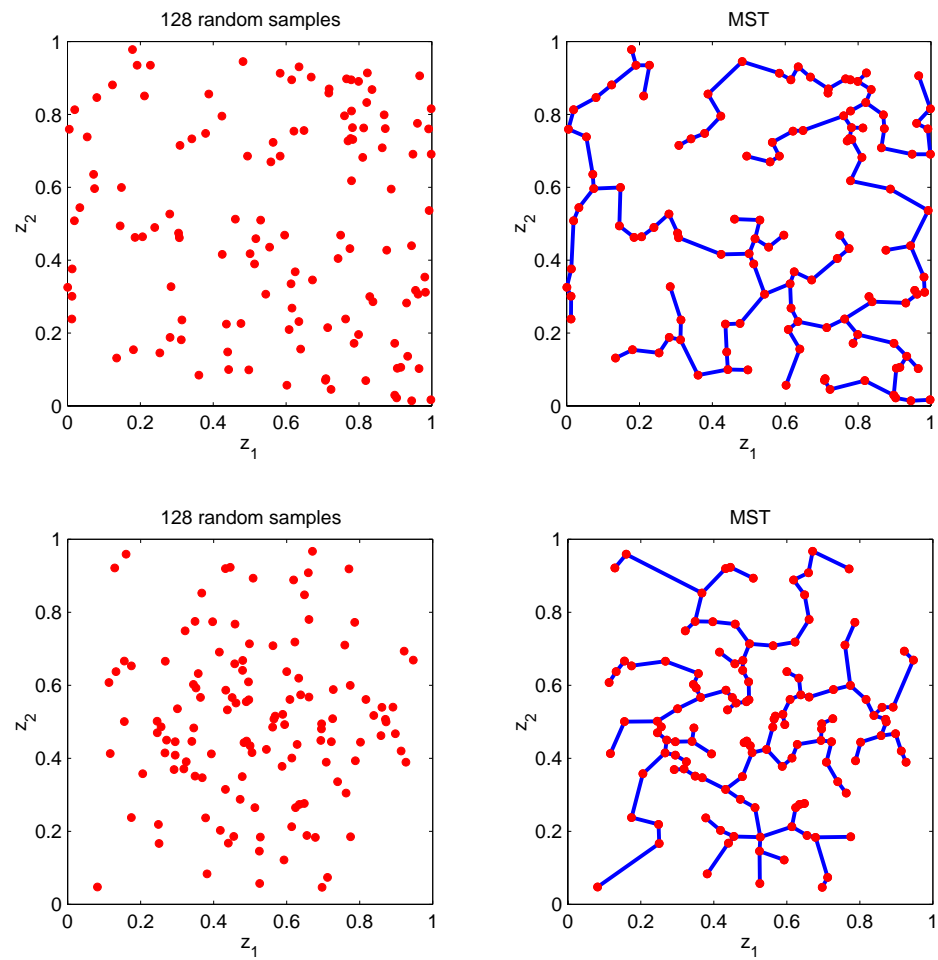


Figure 1:

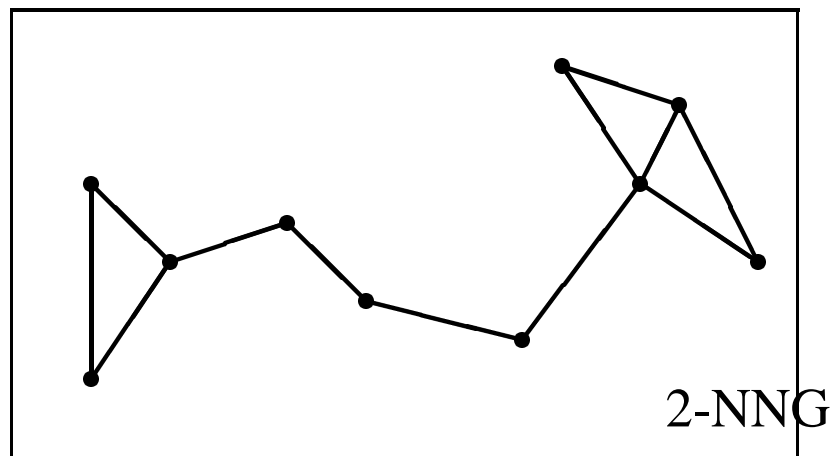
Minimal Euclidean graphs: k -NNG

Example: k -Nearest Neighbors Graph (k -NNG)

Let $N_{k,i}(X_n)$ denote the possible sets of k edges connecting point x_i to all other points in X_n .

The Euclidean Power Weighted k -NNG is

$$L_\gamma^{k-NNG}(X_n) = \sum_{i=1}^n \min_{N_{k,i}(X_n)} \sum_{e \in N_{k,i}(X_n)} |e|^\gamma$$



Large n behavior of MST

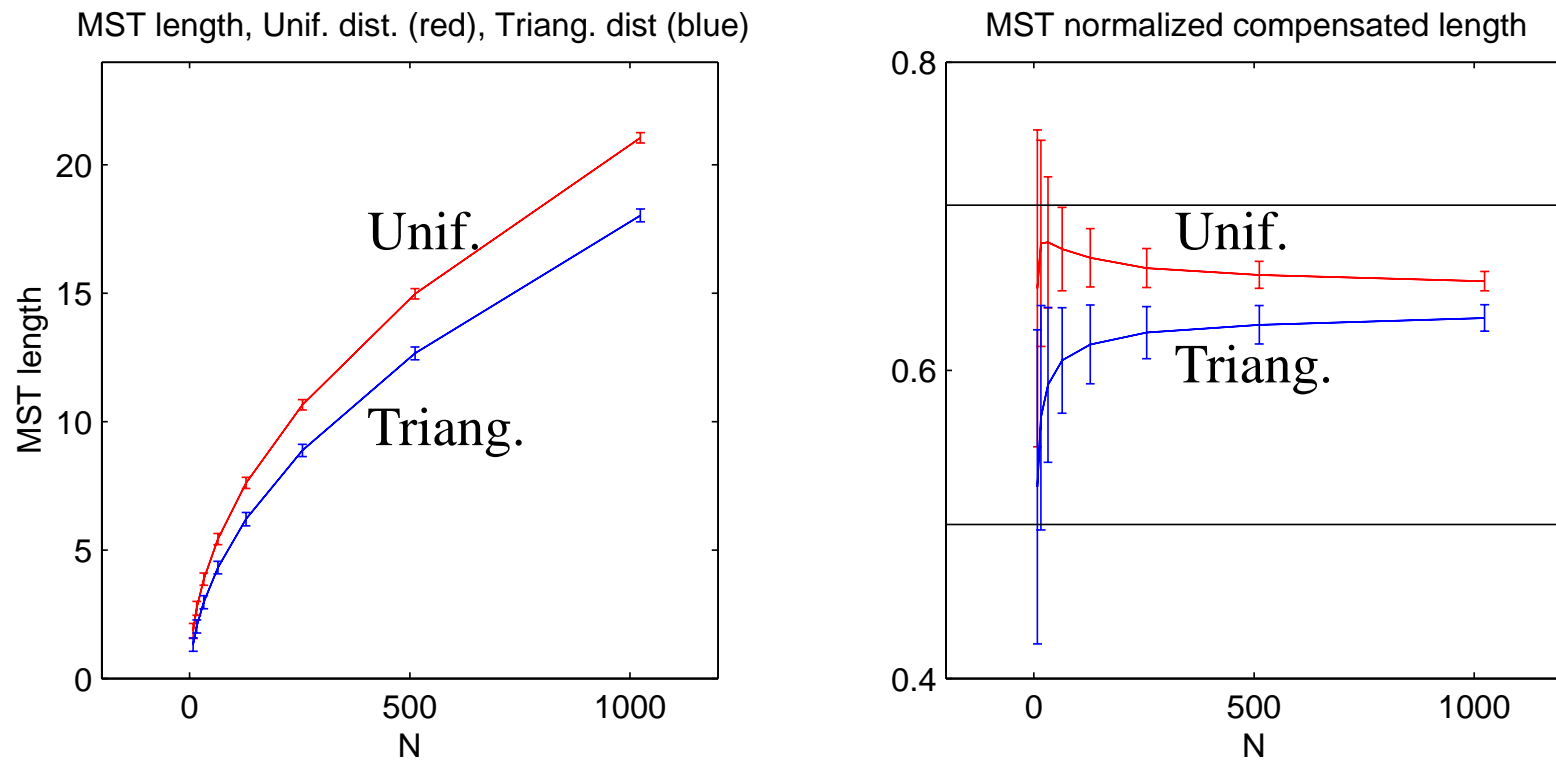


Figure: MST and log MST weights as function of the number of samples.

Asymptotics: the BHH Theorem

Define the MST length functional

$$L_\gamma(X_n) = \min_{T_n} \sum_{e \in T_n} \|e\|^\gamma.$$

Theorem 1 (Beardwood&etal:Camb59) *Let $X_n = \{X_1, \dots, X_n\}$ be an i.i.d. realization from a Lebesgue density f on $[0, 1]^d$.*

$$\lim_{n \rightarrow \infty} L_\gamma(X_n) / n^{(d-\gamma)/d} = \beta_{L_\gamma, d} \int f(x)^{(d-\gamma)/d} dx, \quad (a.s.)$$

Or, letting $\alpha = (d - \gamma) / d$

$$\lim_{n \rightarrow \infty} L_\gamma(X_n) / n^\alpha = \beta_{L_\gamma, d} \exp((1 - \alpha)H_\alpha(f)), \quad (a.s.)$$

Question: What is r.m.s. rate of convergence?

Find constant r such that

$$E^{1/2} \left[\left| L_\gamma(X_n) / n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int f(x)^{(d-\gamma)/d} dx \right|^2 \right] \leq O(n^{-r})$$

Method: adopt Yukich's general setting of quasi-additive continuous Euclidean functionals

Quasi-additive continuous Euclidean functionals

L_γ is a Euclidean functional over \mathbf{R}^d if for every finite subset F of $[0, 1]^d$

$$\forall \mathbf{y} \in \mathbf{R}^d, L_\gamma(F + \mathbf{y}) = L_\gamma(F), \quad (\text{translation invariance})$$

$$\forall c > 0, L_\gamma(cF) = c^\gamma L_\gamma(F), \quad (\text{homogeneity})$$

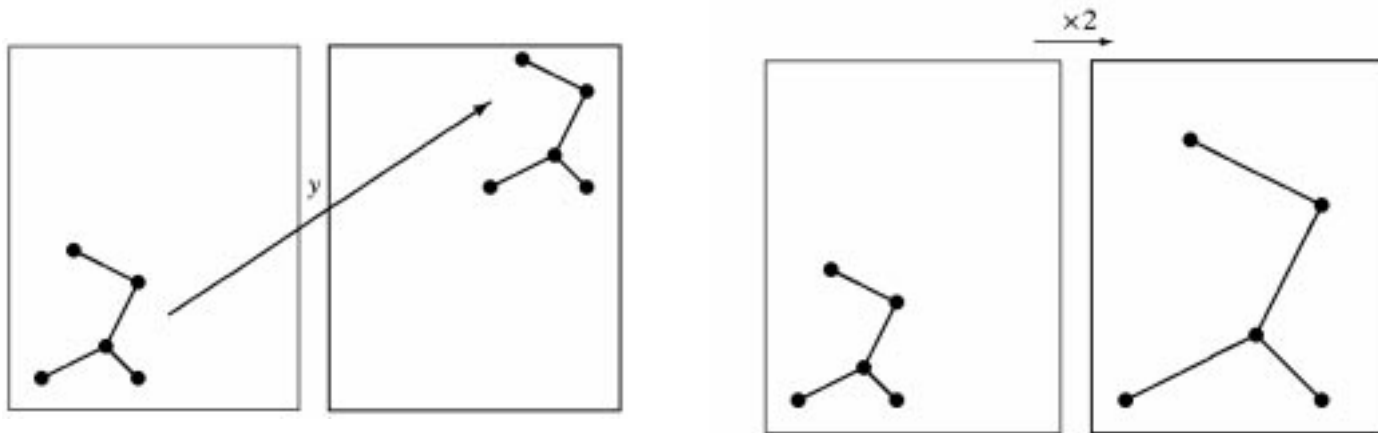


Figure 2: Translation invariance and homogeneity

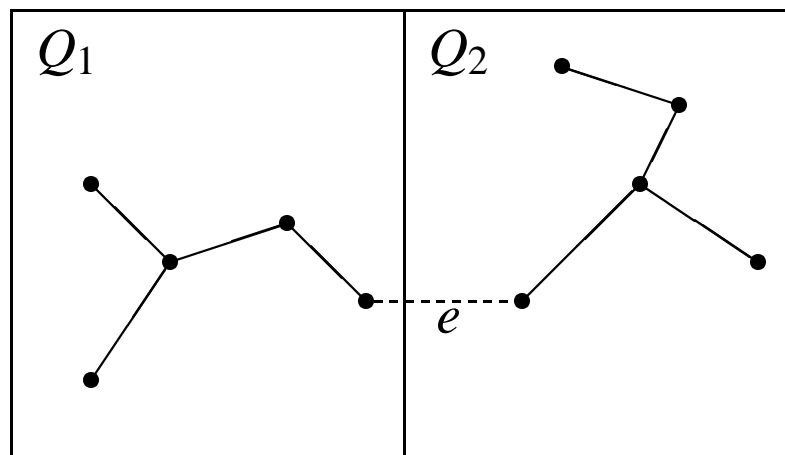
Quasi-additive continuous Euclidean functionals

Let L_γ be a Euclidean functional. Define

- **Null Condition:** $L_\gamma(\phi) = 0$, where ϕ is the null set.
- **Subadditivity:** There exists a constant C_1 with the following property:

For any uniform resolution $1/m$ -partition Q^m

$$L_\gamma(F) \leq m^{-1} \sum_{i=1}^{m^d} L_\gamma(m[(F \cap Q_i) - q_i]) + C_1 m^{d-\gamma}$$



- **Superadditivity:** For the same conditions as above, there exists a constant C_2 s.t.

$$L_\gamma(F) \geq m^{-1} \sum_{i=1}^{m^d} L_\gamma(m[(F \cap Q_i) - q_i]) - C_2 m^{d-\gamma}$$

- **Continuity:** There exists a constant C_3 such that for all finite subsets F and G of $[0, 1]^d$

$$|L_\gamma(F \cup G) - L_\gamma(F)| \leq C_3 (\text{card}(G))^{(d-\gamma)/d}$$

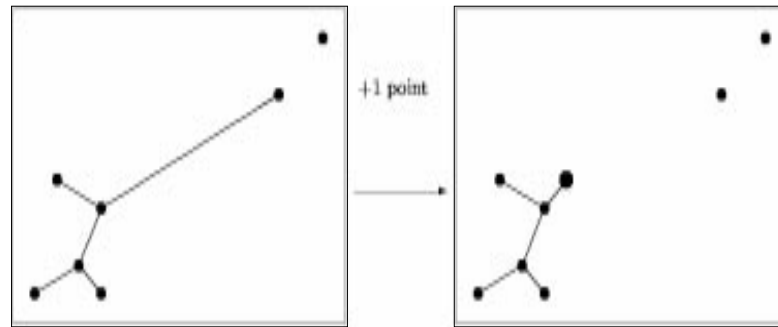


Figure 3: A non-continuous K-MST graph

Definition 1 *A continuous subadditive functional L_γ is said to be a quasi-additive functional when there exists a continuous superadditive functional L_γ^* which satisfies $L_\gamma(F) + 1 \geq L_\gamma^*(F)$ and the approximation property*

$$|E[L_\gamma(U_1, \dots, U_n)] - E[L_\gamma^*(U_1, \dots, U_n)]| \leq C_4 n^{(d-\gamma-1)/d} \quad (2)$$

where U_1, \dots, U_n are i.i.d. uniform random vectors in $[0, 1]^d$.

Another smoothness condition

Definition 2 L_γ is said to satisfy the add-one bound when

$$|E[L_\gamma(U_1, \dots, U_{n+1})] - E[L_\gamma(U_1, \dots, U_n)]| \leq C_4 n^{-\gamma/d} \quad (3)$$

where U_1, \dots, U_{n+1} are i.i.d. uniform random vectors in $[0, 1]^d$.

Convergence rate for uniform f

Theorem 2 (Thm 5.2 Yukich:1998) *Let L_γ be a quasi-additive continuous Euclidean functional which satisfies the add-one bound. Assume that $f(\mathbf{x})$ is uniform over $[0, 1]^d$. Then for all $d \geq 2$ and $1 \leq \gamma < d$*

$$\left| E[L_\gamma(X_n)]/n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int f(x)^{(d-\gamma)/d} dx \right| \leq O(n^{-1/d})$$

Question: How to extend to non-uniform f ?

1. Extend to piecewise constant “block densities” over a uniform partition Q^m :

$$f(x) = \sum_{i=1}^{m^d} \phi_i 1_{Q_i}(x)$$

2. Extend to space of densities sufficiently well approximated by block densities.
3. Obtain worst-case bound on rate over this space of densities.

Block densities

For a set of non-negative constants $\{\phi_i\}_{i=1}^{m^d}$ satisfying $\sum_{i=1}^{m^d} \phi_i = m^d$, define

$$\phi(x) = \sum_{i=1}^{m^d} \phi_i 1_{Q_i}(x)$$

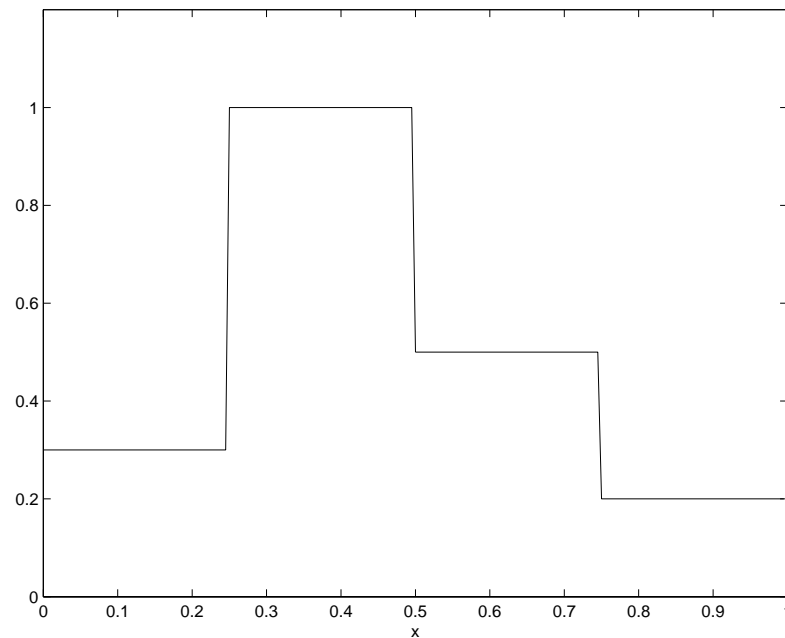


Figure: Block density over the unit interval.

A rate result for block densities

Proposition 1 *Let $d \geq 2$ and $1 \leq \gamma \leq d - 1$. Assume X_1, \dots, X_n are i.i.d. sample points over $[0, 1]^d$ whose marginal is a block density f with m^d levels and support $S \subset [0, 1]^d$. Then for any continuous quasi-additive Euclidean functional L_γ of order γ which satisfies the add-one bound*

$$\left| E[L_\gamma(X_1, \dots, X_n)] / n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int_S f^{(d-\gamma)/d}(x) \, dx \right| \leq O\left((nm^{-d})^{-1/d}\right).$$

Extension to general densities

Define the resolution- m block density approximation of f by

$$\phi(x) = \sum_{i=1}^{m^d} \phi_i 1_{Q_i}(x),$$

where $\phi_i = m^d \int_{Q_i} f(x) dx$.

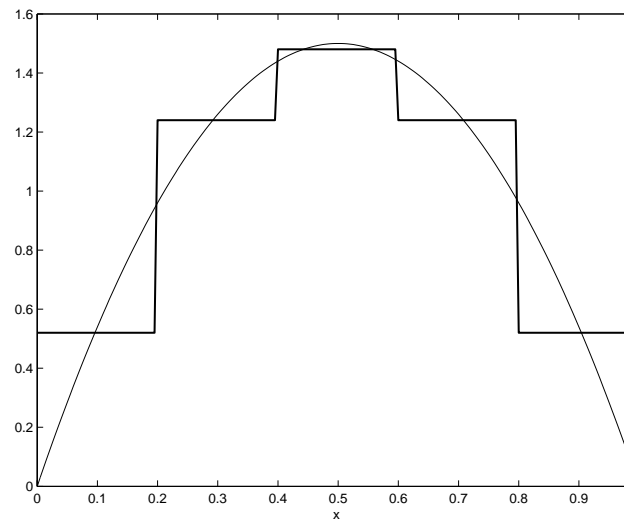


Figure 4: Block density approximation over the unit interval.

Three term bound

By triangle inequality

$$\begin{aligned} & \left| E[L_\gamma(X_1, \dots, X_n)] / n^{\frac{d-\gamma}{d}} - \beta_{L_\gamma, d} \int_S f^{\frac{d-\gamma}{d}}(x) dx \right| \\ & \leq \left| E[L_\gamma(\tilde{X}_1, \dots, \tilde{X}_n)] / n^{\frac{d-\gamma}{d}} - \beta_{L_\gamma, d} \int_S \phi^{\frac{d-\gamma}{d}}(x) dx \right| \quad (I) \\ & \quad + \beta_{L_\gamma, d} \left| \int_S \phi^{\frac{d-\gamma}{d}}(x) dx - \int_S f^{\frac{d-\gamma}{d}}(x) dx \right| \quad (II) \\ & \quad + \left| E[L_\gamma(X_1, \dots, X_n)] - E[L_\gamma(\tilde{X}_1, \dots, \tilde{X}_n)] \right| / n^{\frac{d-\gamma}{d}} \quad (III) \end{aligned}$$

1. Bound on I directly follows from Proposition 1
2. Bound on II is block density approximation error
3. Bound on III is error due to block realizations instead of true realizations of X

Sobolev Spaces

Consider the Sobolev space of L_p functions on \mathbf{R}^d

$$W^{1,p}(\mathbb{R}^d) = L_p(\mathbb{R}^d) \cap \{f : D_{x_j}f \in L_p(\mathbb{R}^d), 1 \leq j \leq d\} .$$

- $D_{x_j}f$ is the x_j -th *weak derivative* of f which satisfies

$$\int_{\mathbb{R}^d} f(x) \frac{\partial}{\partial x_j} \phi(x) dx = - \int_{\mathbb{R}^d} D_{x_j}f(x) \phi(x) dx$$

for any function ϕ infinitely differentiable with compact support.

- $W^{1,p}$ is equipped with a norm

$$\|f\|_{1,p} = \|f\|_p + \|Df\|_p .$$

Approximation Lemma

Lemma 1 *For $1 \leq p < \infty$, let $f \in W^{1,p}(\mathbb{R}^d)$ have support $S \subset [0, 1]^d$. Then there exists a constant $C_6 > 0$, independent of m , such that*

$$\int_S |\phi(x) - f(x)| dx \leq C_6 m^{-\lambda(p)} (\|\mathbf{D}f\|_p + o(1)), \quad (4)$$

where

$$\lambda(p) = \begin{cases} 1, & 1 \leq p \leq d \\ d + 1 - d/p, & d < p < \infty \end{cases}$$

An m -dependent bound

Using the Lemma to bound II and III we obtain

$$\begin{aligned} & \left| E[L_\gamma(X_1, \dots, X_n)] / n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int_S f(x)^{(d-\gamma)/d} dx \right| \\ & \leq \frac{K_1 + C_4}{(nm^{-d})^{1/d}} \left(\int_S f^{\frac{d-1-\gamma}{d}}(x) dx + o(1) \right) \\ & \quad + \frac{\beta_{L_\gamma, d}}{(nm^{-d})^{1/2}} \left(\int_S f^{\frac{1}{2} - \frac{\gamma}{d}}(x) dx + o(1) \right) \\ & \quad + \frac{K_2}{(nm^{-d})^{(d-\gamma)/d}} \\ & \quad + (\beta_{L_\gamma, d} + C'_3) C'_6 m^{-\lambda(p)(d-\gamma)/d} \left(\|Df\|_p^{(d-\gamma)/d} + o(1) \right) \\ & \quad + \frac{1}{n^{(d-\gamma)/d}} \end{aligned}$$

An m -independent bound

By selecting m as the function of n that minimizes this bound, we obtain

Proposition 2 *Let $d \geq 2$ and $1 \leq \gamma \leq d - 1$. Assume X_1, \dots, X_n are i.i.d. random vectors over $[0, 1]^d$ with density $f \in W^{1,p}(\mathbb{R}^d)$, $1 \leq p < \infty$, having support $S \subset [0, 1]^d$. Assume also that $f^{\frac{1}{2} - \frac{\gamma}{d}}$ is integrable over S . Then, for any continuous quasi-additive Euclidean functional L_γ of order γ that satisfies the add-one bound*

$$\left| E[L_\gamma(X_1, \dots, X_n)] / n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int_S f^{(d-\gamma)/d}(x) dx \right| \leq O\left(n^{-r_1(d, \gamma, p)}\right),$$

where

$$r_1(d, \gamma, p) = \frac{\alpha \lambda(p)}{\alpha \lambda(p) + 1} \frac{1}{d}$$

and $\alpha = \frac{d-\gamma}{d}$ and $\lambda(p)$ is defined in Lemma 1.

Concentration inequality

Rhee's inequality for Euclidean functionals (AnnAppProb:93):

$$P\left(\left|L_\gamma(X_1, \dots, X_n) - E[L_\gamma(X_1, \dots, X_n)]\right| > t\right) \leq C \exp\left(\frac{-(t/C_3)^{2d/(d-\gamma)}}{Cn}\right),$$

Hence,

$$\begin{aligned} E\left[\left|L_\gamma(X_1, \dots, X_n) - E[L_\gamma(X_1, \dots, X_n)]\right|^\kappa\right] \\ &= \int_0^\infty P\left(\left|L_\gamma(X_1, \dots, X_n) - E[L_\gamma(X_1, \dots, X_n)]\right| > t^{1/\kappa}\right) dt \\ &\leq C_3 C \int_0^\infty \exp\left(\frac{-t^{2d/[\kappa(d-\gamma)]}}{Cn}\right) dt \\ &= A_\kappa n^{\kappa(d-\gamma)/(2d)} \end{aligned}$$

Main convergence result

Corollary 1 *Let $d \geq 2$ and $1 \leq \gamma \leq d - 1$. Assume X_1, \dots, X_n are i.i.d. random vectors over $[0, 1]^d$ with density $f \in W^{1,p}(\mathbb{R}^d)$, $1 \leq p < \infty$, having support $S \subset [0, 1]^d$. Assume also that $f^{\frac{1}{2} - \frac{\gamma}{d}}$ is integrable over S . Then, for any continuous quasi-additive Euclidean functional L_γ of order γ that satisfies the add-one bound*

$$E \left[\left| L_\gamma(X_1, \dots, X_n) / n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int_S f^{(d-\gamma)/d}(x) dx \right|^\kappa \right]^{1/\kappa} \leq O \left(n^{-r_1(d, \gamma, p)} \right),$$

where

$$r_1(d, \gamma, p) = \begin{cases} \frac{\alpha}{d(\alpha+1)} & , \quad 1 \leq p \leq d \\ \frac{\alpha}{d(\alpha + \frac{1}{d+1-d/p})} & , \quad d < p < \infty \end{cases}$$

where $\alpha = \frac{d-\gamma}{d}$.

Extension to partition approximations

$$L_{\gamma}^m(X_n) = \sum_{i=1}^{m^d} L_{\gamma}(X_n \cap Q_i) + b(m),$$

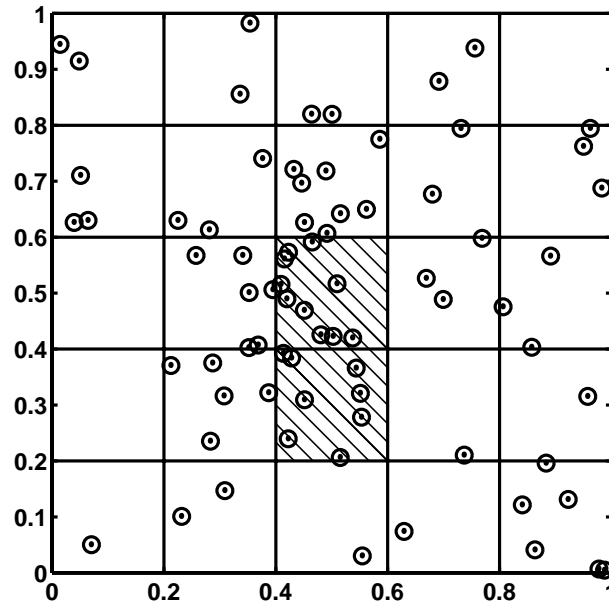


Figure 5: *Partition approximation.*

Pointwise closeness bound

Two Euclidean functionals L_γ and L_γ^* are said to satisfy the *pointwise closeness bound* if

$$|L_\gamma(F) - L_\gamma^*(F)| \leq \begin{cases} C[\text{card}(F)]^{(d-\gamma-1)/(d-1)}, & 1 \leq \gamma < d-1 \\ C \log \text{card}(F), & \gamma = d-1 \neq 1 \\ C, & d-1 < \gamma < d \end{cases},$$

for any finite $F \subset [0, 1]^d$. This condition is satisfied by the MST, TSP and minimal matching function (Lemma 3.7 Yukich:98).

Corollary 2 *Let $d \geq 2$ and $1 \leq \gamma < d - 1$. Assume that the Lebesgue density $f \in W^{1,p}(\mathbb{R}^d)$, $1 \leq p < \infty$ has support $S \subset [0, 1]^d$. Assume also that $f^{1/2-\gamma/d}$ is integrable over S . Let $L_\gamma^m(X_n)$ be a partition approximation to $L_\gamma(X_n)$ where L_γ is a continuous quasi-additive functional of order γ which satisfies the pointwise closeness bound and the add-one bound. Then if $b(m) = O(m^{d-\gamma})$*

$$E \left[\left| L_\gamma^{m(n)}(X_1, \dots, X_n) / n^{(d-\gamma)/d} - \beta_{L_\gamma, d} \int_S f^{(d-\gamma)/d}(x) dx \right| \right] \leq O \left(n^{-r_2(d, \gamma, p)} \right), \quad (5)$$

where

$$r_2(d, \gamma, p) = \frac{\alpha \lambda(p)}{\frac{d-1}{\gamma} \alpha \lambda(p) + 1} \frac{1}{d},$$

where $\alpha = \frac{d-\gamma}{d}$ and $\lambda(p)$ is defined in Lemma 1. This rate is attained by choosing the progressive-resolution sequence

$$m = m(n) = n^{1/[d(\frac{d-1}{\gamma} \alpha \lambda(p) + 1)]}.$$

Application to Entropy Estimation

Consider entropy estimates

$$\hat{H}_\alpha = (1 - \alpha)^{-1} \log \hat{I}_\alpha,$$

where \hat{I}_α is a consistent estimator of the integral

$$I_\alpha(f) = \int f^\alpha(x) dx$$

- $\hat{I}_\alpha = L_\gamma(X_1, \dots, X_n) / (\beta_{L_\gamma, d} n^\alpha)$ is a strongly consistent estimator of $I_\alpha(f)$;
- indirect estimator: given non-parametric function estimates \hat{f} of f define the function plug-in estimator $\hat{I}_\alpha = I_\alpha(\hat{f})$.

Convergence rate comparisons

Question: How fast is

$$|E[\hat{H}_\alpha] - H_\alpha(f)| \rightarrow 0, \text{ when } n \rightarrow \infty?$$

Find $r > 0$ such that

$$|E[\hat{H}_\alpha] - H_\alpha(f)| = O(n^{-r})$$

For both cases

$$|\hat{H}_\alpha - H_\alpha(f)| = \frac{1}{1-\alpha} \frac{|\hat{I}_\alpha - I_\alpha(f)|}{I_\alpha(f)} + o(|\hat{I}_\alpha - I_\alpha(f)|)$$

\Rightarrow convergence rate of $|E[\hat{H}_\alpha] - H_\alpha(f)|$ is identical to that of $|E[\hat{I}_\alpha] - I_\alpha(f)|$.

Multivariate Besov function space

Definition 3 (Nikolskii:75) Let $1 \leq p, q < \infty$, $\sigma > 0$ and k, ρ be nonnegative integers satisfying the inequalities $k > \sigma - \rho > 0$. The function f belongs to the class $B_{p,q}^\sigma(\mathbb{R}^d)$ if $f \in L_p(\mathbb{R}^d)$ and there exist partial weak derivatives $D^{(s)} f = \partial^\rho f / \partial x_1^{s_1} \dots \partial x_d^{s_d}$ of order $s = (s_1, \dots, s_d)$ ($|s| = s_1 + \dots + s_d = \rho$) such that the following seminorm is finite:

$$\|f\|_{b_{p,q}^\sigma} = \sum_{|s|=\rho} \left\{ \int_{\mathbb{R}^d} \left(\frac{\|\Delta_\tau^k D^{(s)} f\|_p}{|\tau|^{\sigma-\rho}} \right)^q \frac{d\tau}{|\tau|^d} \right\},$$

where Δ_τ^k is an operator which takes the k -th order finite difference in the direction of τ .

Sobolev vs Besov

Lemma 2 (*Besov&etal:79*) *Let $p > d$ and let σ be a positive integer. Then*

$$B_{p,1}^{\sigma}(\mathbb{R}^d) \subset W^{\sigma,p}(\mathbb{R}^d)$$

Asymptotic convergence of entropy estimators

Proposition 3 *Let $p > d \geq 2$ and $\alpha = (d - \gamma)/d \in [1/2, (d - 1)/d]$*

$$\sup_{f^\alpha \in B_p^{1,1}} E^{1/\kappa} \left[\left| \int \widehat{f}^\alpha(x) dx - \int f^\alpha(x) dx \right|^\kappa \right] \geq O \left(n^{-1/(2+d)} \right)$$

while,

$$\sup_{f^\alpha \in B_p^{1,1}} E^{1/\kappa} \left[\left| \frac{L_\gamma(X_1, \dots, X_n)}{n^\alpha} - \beta_{L_\gamma, d} \int f^\alpha(x) dx \right|^\kappa \right] \leq O \left(n^{-\frac{\alpha \lambda(p)}{1 + \alpha \lambda(p)} \frac{1}{d}} \right)$$

where $\lambda(p) = d + 1 - d/p$.

Note: minimal-graph estimator converges faster for

$$\alpha \geq \frac{1}{2} \frac{d}{d + 1 - d/p}$$

Open problems

- ▷ Extend convergence rates to:
 - smoother densities, e.g. in $W^{\sigma,p}(\mathbf{R}^d)$
 - densities with unbounded support.
 - weaker quasi-additive continuity conditions
- ▷ Analysis of clustering algorithms in entropic graph setting.
- ▷ Analysis of distributional properties of entropic graphs.