# Local Nearest Neighbor (LNN) Information Estimator

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#### 1 Introduction

This document is a description of the Python package for Local Nearest Neighbor (LNN) Information estimator. You can download the code from http://github.com/liverlover/lnn/. The paper describing LNN estimators can be found in [1]. The repository contains the following files:

- lnn.py: Main code.
- demo.py: An example of usage.
- readme.pdf: Readme document.

### 2 Functions

#### 2.1 Main Functions

Here are the proposed entropy and mutual information estimator in [1]:

- 1. **entropy(data)**: Estimate the differential entropy H(X) of  $X \in \mathbb{R}^{d_x}$  from samples  $\{x_i\}_{i=1}^N$  using LNN entropy estimator [1, Section 3].
  - data: A 2D list of dimension  $N \times d_x$ , where each row is one sample  $x_i \in \mathbb{R}^{d_x}$ .
  - Output: Scalar  $\widehat{H}(X)$ .
- 2. **mi(data,split)**: Estimate the mutual information I(X;Y) of  $X \in \mathbb{R}^{d_x}$  and  $Y \in \mathbb{R}^{d_y}$  from samples  $\{x_i, y_i\}_{i=1}^N$ , using 3LNN mutual information estimator [1, Section 5].
  - data: A 2D list of dimension  $N \times (d_x + d_y)$ , where each row is one pair of sample  $(x_i, y_i) \in \mathbb{R}^{d_x + d_y}$ .
  - split: Equals to  $d_x$ , telling which part of data represent X and which part represent Y.
  - Output: Scalar  $\widehat{I}(X;Y)$ .

#### 2.2 Other Estimators for Comparison

Here we also provide other entropy and mutual information estimators which are used for comparison in the experiments in [1].

- Entropy Estimators:
  - 1. **KDE\_entropy(data):** KDE entropy estimator [2].
  - 2. **KL\_entropy(data):** Kozachenko-Leonenko entropy estimator [3].
  - 3. LNN\_1\_entropy(data): LNN entropy estimator with order parameter p = 1.
- Mutual Information Estimators:
  - 1. \_3KDE\_mi(data,split): Combination of 3 KDE entropy estimators.

- 2. \_3KL\_mi(data,split): Combinations of 3 Kozachenko-Leonenko entropy estimator.
- 3. \_KSG\_mi(data,split): KSG mutual information estimator [4].
- 4.  $_3$ LNN $_1$ -mi(data,split): Combination of 3 LNN entropy estimators with p = 1.
- 5.  $_3$ LNN $_1$ \_KSG $_mi(data,split)$ : Combinations of 3 LNN entropy estimators with p = 1, using "KSG trick" (see [1, Section 5]).
- 6. \_3LNN\_2\_KSG\_mi(data,split): Combinations of 3 LNN entropy estimators with p = 2, using "KSG trick" (see [1, Section 5]).

# 3 Usage

Here we provide a simple sample of usage of the package. Here X and Y are joint standard Gaussian random variable with high correlation.

```
>> import numpy.random as nr
>> from math import log, pi, exp
>> import lnn
>> r = 0.9999
>> data = nr.multivariate_normal([0,0],[[1,r],[r,1]],100)
>> print "Ground Truth = ", log(2*pi*exp(1))+0.5*log(1-r**2)
Ground Truth = -1.42074452992
>> print "LNN: H(X) = ", lnn.entropy(data)
LNN: H(X) = -1.38841327294
```

You can find the full version of the code in demo.py.

### References

- [1] Gao, W., Oh, S. and Viswanath, P. Breaking the Bandwidth Barrier: Geometrical Adaptive Entropy Estimation, Conference on Neural Information Processing Systems (NIPS), 2016.
- [2] Beirlant J, Dudewicz E J, Györfi L, et al. Nonparametric entropy estimation: An overview[J]. International Journal of Mathematical and Statistical Sciences, 1997, 6(1): 17-39.
- [3] L. F. Kozachenko and N. N. Leonenko. Sample estimate of the entropy of a random vector. Problem Peredachi Informatsii, 23(2):916, 1987.
- [4] Kraskov A, Stögbauer H, Grassberger P. Estimating mutual information[J]. Physical review E, 2004, 69(6): 066138.