



Lossy Information Exchange and Instantaneous Communication

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Final Report

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Final Report on AFOSR Project “Lossy Information Exchange and Instantaneous Communications”, FA9550-11-1-0168

Lizhong Zheng, September 2015

The goal of this project was to develop a new framework to study information exchanges with constraints in the delays and computational complexity, from an information theoretic perspective.

The project is based on a key difference between digital communication problems and statistical inference problems, namely, in communications, we often have the goal of delivering the information as an entirety, such that every bit of the information is reliably conveyed; whereas in inference problems, it is often the case that we are only interested in a special feature in the data, and do not wish to reconstruct the entire observed data after processing. This is a fundamental difference, as in the later cases, one has to distinguish between different components in an observation, make sure the relevant/important information remain intact after processing and transmission, and often discard the rest of the information. This notion of lossy processing, or information discipation in the processing, is a concept that is lacking the conventional information theoretic analysis.

We have, in the first a few years of this project, developed a new framework which we call “linear information coupling”. In this setup, we think any information exchange as a variation of the posterior distribution of the information-carrying message. While the variation of the distribution is often a very high-dimensional object, we define an ortho-normal basis in the space of probability distributions. We choose this basis in correspondence to the SVD spectrum of the observation model, so that we can decompose the calculation of the posterior distributions into computing a sequence of scores. These scores are ordered and labeled by the information contents they contain. With this general structure, we can, instead of computing the entire posterior distribution of the message, compute only a few, say, k , scores from the top of the list, knowing that these score values contains the maximum amount of useful information among all k -dimensional statistics that can be extracted from the observations. We call these scores “efficient statistics”. They are not sufficient statistics in the conventional sense, as they do not contain all the information in the entire observation, but are the most informative functions one can compute, given the computation complexity that one can afford.

The idea of information coupling has been applied to a variety of problems. This includes new techniques that can be used for the traditional network communication problems, as well as inference problems. Some of these results are reported in the following publications.

Shao-Lun Huang, Anuran Makur, Fabian Kozynski, Lizhong Zheng: Efficient Statistics: extracting Information From IID Observations. Allerton Conference 2014

Shao-Lun Huang, Lizhong Zheng: The Linear Information Coupling Problems. CoRR abs/1406.2834

Shao-Lun Huang, Changho Suh, Lizhong Zheng: Euclidean information theory of networks. Accepted IEEE Trans. Info. Theory, August, 2015.

Near the end of this project, we have made another important finding. In all of our previous works, we assumed that the noisy observation channel model is precisely known. This is, however, a problematic assumption in reality. With the help of the SVD structure, we came up with an algorithm, which is a generalization of the Alternating Conditional Expectation (ACE) algorithm, which efficiently estimate the top singular vectors of the DTM, and use them as the score functions, or efficient statistics. This put our result in the general category of dimension reduction/non-linear feature selection problems.

Theoretically, the reason for this simple algorithm to work is related to the concept of Renyi maximal correlation. In 1959, Renyi proposed a measure of the dependence between two random variables, by finding a pair of functions, one for each of the random variables, which are highly correlated. In our application, one variable corresponds to the user, the other to the service. Finding the Renyi correlation in this problem can thus be understood as finding a particular aspect of the services that can be used to distinguish the users, i.e. to answer the question 'which user is more likely to use a certain portfolio of services?' Our algorithm is in fact a very efficient way, not only to compute the maximal Renyi correlation, but also to find the best choices of 200 pairs of functions. As we distill everything we know about a user into a 200 dimensional signature, we inevitably discard some knowledge about this user; but with our algorithm, it is guaranteed that these are the 200 values that carry the most amount of useful information, for the purpose of predicting their service preferences.

The conceptual step made in this algorithm is that instead of estimating the complete statistical model of the observation channel, we can instead only estimate one "mode" of it, corresponding to the maximal Renyi correlation, and the singular vector of the DTM with the leading singular value. We show that this requires a significantly smaller number of training samples, which is the critical issue for most Big Data problems. On the other hand, the resulting statistics are also information theoretically optimal, in the sense that they carry the largest amount of information, for the given number of statistics. It is thus quite surprising that our result gives the optimal tradeoff between three objectives: the inference performance, the number of statistics used, and the sample complexity to learn these score functions.

There are some further advantages of our algorithm. The most useful one in practice is the generality of the algorithm. Instead of requiring specific forms of data, and

restricting to specific type of score functions, such as linear functions over real-valued data in PCA, our algorithm is much more general, and can be used to process information with different types of data, discrete or continuous valued. The score functions selected from this approach are in general non-linear, which shed lights to the difficult problem of non-linear feature selection. We can also put additional constraints for specific types of score functions in our formulation. For example, when we apply to real valued data and insist the score functions to be linear, we indeed recover the PCA solution as a special case. Thus, our algorithm can be used in process and integrate data from different applications, to jointly improve the performance in inference.

At the very end of this project, we just started reporting our results in publications. The first one of them is

Anuran Makur, Shao-Lun Huang, Fabian Kozynski, Lizhong Zheng, “An Efficient Algorithm for Information Decomposition and Extraction” Allerton conference, Oct. 2015

We have applied this algorithm in a number of realistic problems, including the Netflix problem, a community detection problem with Facebook connection graphs, and a detection problem with audio signals.

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Lizhong Zheng

Program Manager

The AFOSR Program Manager currently assigned to the award

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Abstract

In this project, we studied a new theoretical framework for information transmission with the focus on extracting only a part of the information. This formulation is particularly useful when we apply information theory to data analysis problems, where the goal is different from the full and reliable information recovery needed in the classical communications problems. We developed a geometric structure for the space of probability distributions, and a new method to decompose the information carried by the observed data based on that. This formulation gives a general setup to understand dimension reduction as lossy information processing procedures and gives a new operational meaning of information metrics, in the context of data analytics. We can quantitatively describe the information efficiency, computation complexity, and sample complexity to learn the model in one picture. We also make connections to the existing results on dimensional reduction. Based on this framework, we developed new algorithms for dimension reduction and non-linear feature selection. We proved a number of optimality results for these new constructions, and applied the algorithm in real data analysis tasks.

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