

MAXENTMC – A MAXIMUM ENTROPY ALGORITHM WITH MOMENT CONSTRAINTS

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1. INTRODUCTION

To be written.

2. MATHEMATICAL FORMULATION OF THE MAXIMUM ENTROPY PROBLEM WITH MOMENT CONSTRAINTS

The content of this section is based on my paper titled “*The multidimensional maximum entropy moment problem: A review on numerical methods*”, published in Commun. Math. Sci. Some parts of the section content below may be copied, with minor changes, from that work. For more details, please refer to that work and references therein.

2.1. Notations. Let N be a positive integer number. Let \mathbb{R}^N be an N -dimensional space of real numbers, and let \mathbb{Z}^N be an N -dimensional space of nonnegative integer numbers. Let x denote an element of \mathbb{R}^N , and let i denote an element of \mathbb{Z}^N . We define a *monomial* x^i as

$$(2.1) \quad x^i = x_1^{i_1} x_2^{i_2} \dots x_N^{i_N} = \prod_{k=1}^N x_k^{i_k}.$$

We denote the *power* $|i|$ of the monomial in (2.1) as

$$(2.2) \quad |i| = i_1 + i_2 + \dots + i_N = \sum_{k=1}^N i_k.$$

Let U be a domain in \mathbb{R}^N . Let $\rho : U \rightarrow \mathbb{R}$ denote any nonnegative continuous function on U , with the condition that

$$(2.3) \quad \int_U \rho(x) dx < \infty.$$

In this case, ρ is called the *density*. Given an element $\mathbf{i} \in \mathbb{Z}^N$, we denote the corresponding *moment* $m_{\mathbf{i}}[\rho]$ as

$$(2.4) \quad m_{\mathbf{i}}[\rho] = \int_U \mathbf{x}^{\mathbf{i}} \rho(x) dx.$$

Let P be a finite subset of \mathbb{Z}^N . Let $\rho_c : U \rightarrow \mathbb{R}$ denote any ρ for which all $m_{\mathbf{i}}$, such that $\mathbf{i} \in P$, are finite, and are given by a set of real numbers $c_{\mathbf{i}} \in \mathbb{R}$, for all $\mathbf{i} \in P$:

$$(2.5) \quad m_{\mathbf{i}}[\rho_c] = c_{\mathbf{i}}, \quad \forall \mathbf{i} \in P.$$

It is then said that ρ_c possesses a set of moments (or *moment constraints*, in the context of what is presented below) c . There are two things which need to be emphasized:

- (1) For an arbitrarily specified set of real numbers c , the corresponding density ρ_c does not have to exist (for example, suppose that \mathbf{i} includes only even integers, yet $c_{\mathbf{i}} < 0$);
- (2) When ρ_c nonetheless exists, it does not have to be unique.

What is presented below deals with the second situation.

2.2. Maximum entropy under moment constraints. We define the *Shannon entropy* $S[\rho]$ as

$$(2.6) \quad S[\rho] = - \int_U \rho(x) \ln \rho(x) dx.$$

Shannon entropy is recognized as a measure of uncertainty in probability densities. In the context of the maximum entropy problem, the goal is to find ρ_c^* among all ρ_c , such that it maximizes S :

$$(2.7) \quad \rho_c^* = \arg \max_{\rho_c} S[\rho_c].$$

It can be shown via variational analysis that ρ_c^* belongs to the family of functions $f_{\lambda} : U \rightarrow \mathbb{R}$, where $\lambda = \{\lambda_{\mathbf{i}}\}$ are real numbers (called the *Lagrange multipliers*), $\mathbf{i} \in P$. $f_{\lambda}(x)$ is explicitly given by

$$(2.8) \quad f_{\lambda}(x) = \exp \left(\sum_{\mathbf{i} \in P} \lambda_{\mathbf{i}} x^{\mathbf{i}} \right).$$

Depending on the sets U , P and on the values λ , $f_{\lambda}(x)$ may or may not be a density; indeed, observe that while any f_{λ} is certainly nonnegative, its integral over U does not have to be finite. More precisely, any f_{λ} is a density over a finite domain U , but not all f_{λ} are densities when U is not a finite domain.

Our goal now is to find the set λ^* for which

$$(2.9) \quad f_{\lambda^*}(x) = \rho_c^*(x), \quad \forall x \in U.$$

This is accomplished by computing the minimum of the *Lagrangian function* (or shortly, the Lagrangian)

$$(2.10) \quad L(\lambda) = \int_U f_\lambda(x) \, dx - \sum_{i \in P} c_i \lambda_i.$$

The gradient (the vector of the first derivatives) and hessian (the matrix of the second derivatives) of the Lagrangian are given, respectively, via

$$(2.11a) \quad \frac{\partial L}{\partial \lambda_i} = m_i[f_\lambda] - c_i,$$

$$(2.11b) \quad \frac{\partial^2 L}{\partial \lambda_i \partial \lambda_j} = m_{i+j}[f_\lambda].$$

From the above expressions, it is easy to see that the gradient of L is zero when the moment constraints are met (that is, a critical point of L is found), and that the hessian of L is positive definite, which means that the critical point is indeed a unique minimum. Note that, in general, the critical point does not have to exist.

3. HOW TO SPECIFY CONSTRAINTS AND LAGRANGE MULTIPLIERS

To be written.

4. HOW TO COMPUTE MOMENTS

To be written.

5. HOW TO COMPUTE THE GRADIENT AND HESSIAN OF THE OBJECTIVE FUNCTION

To be written.