

The MMSE Conjecture

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Abstract

This note presents the MMSE conjecture to attract possible interest.

Let f be a probability density on the real line. Provided there are well-defined, consider the entropy of f ,

$$H(f) = \int_{\mathbb{R}} f \log f \, dx$$

(up to the sign convention) and its Fisher information

$$I(f) = \int_{\mathbb{R}} \frac{f'^2}{f} \, dx.$$

Denote by $p_t(x)$ the Gaussian kernel

$$p_t(x) = \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t}, \quad t > 0, \quad x \in \mathbb{R},$$

and let $f_t = f * p_t$, $t > 0$.

The MMSE conjecture basically asks whether the knowledge of $H(f_t)$, $t > 0$, or $I(f_t)$, $t > 0$, is enough to determine f .

To precisely describe the conjecture (and the terminology), let us slightly extend the picture.

Let μ be a probability measure on the Borel sets of \mathbb{R} . For every $t > 0$, let μ_t be the convolution of μ with the Gaussian kernel p_t . That is, μ_t has the strictly positive C^∞ density f_t with respect to the Lebesgue measure which is given as the convolution

$$f_t(x) = \int_{\mathbb{R}} p_t(x-y) d\mu(y), \quad x \in \mathbb{R}.$$

From a probabilistic perspective, let X be a random variable on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ with law μ . Then μ_t is the law of $X_t = X + \sqrt{t}N$ where N is a standard normal variable independent of X .

Now, $I(f_t)$ is well-defined for every $t > 0$ (see below) and we may ask whether its knowledge determines μ . The classical de Bruijn formula expresses that

$$\frac{d}{dt} H(f_t) = -\frac{1}{2} I(f_t)$$

so that the question may be formulated equivalently for the entropy.

In the sequel, write $I(\mu_t) = I(X_t) = I(f_t)$, $t > 0$.

It is immediate to see that the quantities $I(\mu_t)$, $t > 0$, are invariant by translation and symmetry. That is, if ν is the distribution defined by

$$\int_{\mathbb{R}} \varphi d\nu = \int_{\mathbb{R}} \varphi(\pm x + a) d\mu(x)$$

for every bounded measurable $\varphi : \mathbb{R} \rightarrow \mathbb{R}$, where $a \in \mathbb{R}$, then $I(\nu_t) = I(\mu_t)$ for every $t > 0$. Equivalently, $I(X_t)$, $t > 0$, is invariant by the change of X into $\pm X + a$ for any $a \in \mathbb{R}$.

Conjecture. *Let μ be a probability distribution on \mathbb{R} . Then the knowledge of $I(\mu_t)$, $t > 0$, characterizes μ up to translation and symmetry.¹*

In probabilistic terms, does $I(X_t)$, $t > 0$, characterizes the law of X up to the change of X into $X + a$ or $-X$?

This conjecture came up in information theory in the work [2] by D. Guo, Y. Wu, S. Shamai and S. Verdú on the estimation of a random variable from its observation perturbed by a Gaussian noise. More precisely, in the preceding notation, consider

$$\text{MMSE}(t) = \mathbb{E}\left([X - \mathbb{E}(X | X_t)]^2\right), \quad t \geq 0,$$

which is known as the Minimum Mean-Square Error (MMSE). Note that although $\mathbb{E}(X | X_t)$ may not be well-defined if X is not integrable, it makes sense to consider the integrable random variable $X_t - \mathbb{E}(X | X_t)$ which is identified to $\sqrt{t} \mathbb{E}(N | X_t)$ since

$$X_t = \mathbb{E}(X_t | X_t) = \mathbb{E}(X | X_t) + \sqrt{t} \mathbb{E}(N | X_t).$$

In particular, $X - \mathbb{E}(X | X_t)$ makes also sense and has moments of all orders.

The MMSE connects to the Fisher information $I(X_t)$, $t > 0$, along the heat flow via the identity

$$t^2 I(X_t) = t - \text{MMSE}(t), \quad t > 0.$$

(cf. [1, 3]). (Note that the invariances by translation and symmetry are immediate on this representation.)

Some partial results on the conjecture are described in [1, 2, 3]. The multi-dimensional case may also be discussed.

¹When I mentioned this conjecture to a close colleague, I got the following answer: to characterize a probability distribution, there is a convenient tool, the Fourier transform.

References

- [1] D. Guo, S. Shamai, S. Verdú. Mutual information and minimum mean-square error in Gaussian channels. *IEEE Trans. Inform. Theory* 51 (2005), 1261–1282.
- [2] D. Guo, Y. Wu, S. Shamai, S. Verdú. Estimation in Gaussian noise: properties of the minimum mean-square error. *IEEE Trans. Inform. Theory* 57 (2011), 2371–2385.
- [3] M. Ledoux. Heat Flow Derivatives and Minimum Mean-Square Error in Gaussian Noise. *IEEE Trans. Inform. Theory*, to appear (2016).