

# The median of an exponential family and the normal law

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## Abstract

Let  $P$  be a probability on the real line generating a natural exponential family  $(P_t)_{t \in \mathbb{R}}$ . We show that the property that  $t$  is a median of  $P_t$  for all  $t$  characterizes  $P$  as the standard Gaussian law  $N(0, 1)$ .

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## 1 Introduction

Let  $P$  be a probability on the real line and assume that

$$L(t) = \int_{-\infty}^{+\infty} e^{tx} P(dx) < \infty \quad \text{for } t \in \mathbb{R}. \quad (1)$$

Such a probability generates the natural exponential family

$$\mathcal{F}_P = \{P_t(dx) = \frac{e^{tx}}{L(t)} P(dx), \quad t \in \mathbb{R}\}.$$

Then it might happen that the natural parameter  $t$  of  $\mathcal{F}_P$  is always a median of  $P_t$ , in the sense of

$$P_t((-\infty, t)) \leq \frac{1}{2} \leq P_t((-\infty, t]) \quad \text{for } t \in \mathbb{R}. \quad (2)$$

In the sequel we denote by  $\mathcal{P}$  the set of probabilities  $P$  such that (1) and (2) are fulfilled. A noteworthy example of an element of  $\mathcal{P}$  is the standard normal

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distribution  $N(0, 1)$ , for which  $L(t) = e^{t^2/2}$  and  $P_t = N(t, 1)$ . It will turn out that it is the only one. The following preliminary lemmas simplify the study of  $\mathcal{P}$ .

**Lemma 1.** *If  $P \in \mathcal{P}$ , then  $P$  is absolutely continuous with respect to Lebesgue measure. As a consequence, we have equality throughout in (2).*

**Lemma 2.** *If  $P \in \mathcal{P}$ , then its distribution function is strictly increasing.*

If  $P \in \mathcal{P}$ , then Lemma 1 allows us to write

$$P(dx) = g(x)\varphi(x)dx, \quad (3)$$

where  $g$  is some measurable non-negative function and  $\varphi(x) = e^{-x^2/2}/\sqrt{2\pi}$  denotes the standard normal density, and we will show that then  $g(x) = 1$  a.e. to get:

**Theorem 1.**  $\mathcal{P} = \{N(0, 1)\}$ .

The proofs of the above results are contained in Section 2, followed by a conjecture and a further theorem.

## 2 Proofs

**Proof of Lemma 1.** The next paragraph shows that the distribution function of  $P$  is locally Lipschitz, and this implies the claimed absolute continuity, even with a locally bounded density, compare for example Royden and Fitzpatrick (2010, pp. 120–124).

For  $t \in \mathbb{R}$ , multiplying in assumption (2) by  $L(t)$  yields

$$h(t) := \int_{(-\infty, t]} e^{tx} P(dx) \geq \frac{1}{2}L(t) \geq \int_{(-\infty, t)} e^{tx} P(dx) = h(t-). \quad (4)$$

Hence, if  $A > 0$  is given, then for  $s, t$  with  $-A \leq s < t \leq A$ , we get

$$\begin{aligned} P((s, t)) &= \int_{(s, t)} e^{-tx} e^{tx} P(dx) \leq e^{A^2} \int_{(s, t)} e^{tx} P(dx) \\ &= e^{A^2} \left( h(t-) - h(s) + \int_{(-\infty, s]} (e^{sx} - e^{tx}) P(dx) \right) \\ &\leq e^{A^2} \left( \frac{1}{2}(L(t) - L(s)) + (t - s) \int_{\mathbb{R}} |x| e^{A|x|} P(dx) \right) \\ &\leq c_A \cdot (t - s) \end{aligned}$$

for some finite constant  $c_A$ . We have been using (4) and  $|e^u - e^v| \leq |u - v|e^w$  for  $|u|, |v| \leq w$  at the penultimate step. Using assumption (1), we rely at the ultimate step on local Lipschitzness of  $L$ , due to its analyticity, and on finiteness of  $\int_{\mathbb{R}} |x| e^{A|x|} P(dx)$ , .  $\square$

**Proof of Lemma 2.** Assume to the contrary that there exist  $a, b \in \mathbb{R}$  with  $a < b$  and  $P((a, b)) = 0$ . Then, for  $t \in (a, b)$ , Lemma 1 and (2) yield

$$\int_{-\infty}^a e^{tx} P(dx) = \int_{-\infty}^t e^{tx} P(dx) = \int_t^{+\infty} e^{tx} P(dx) = \int_b^{+\infty} e^{tx} P(dx).$$

Thus the two measures  $\mathbf{1}_{(-\infty, a]}(x)P(dx)$  and  $\mathbf{1}_{[b, +\infty)}(x)P(dx)$  have finite and identical Laplace transforms on some non-empty interval. Hence the two measures coincide, and hence  $P$  must be the zero measure, which is absurd.  $\square$

**Proof of Theorem 1.** With the representation (3) for  $P \in \mathcal{P}$ , assumption (2) is rewritten as

$$\int_{-\infty}^t e^{tx - \frac{x^2}{2}} \frac{1}{\sqrt{2\pi}} g(x) dx = \frac{1}{2} \int_{-\infty}^{+\infty} e^{tx - \frac{x^2}{2}} \frac{1}{\sqrt{2\pi}} g(x) dx. \quad (5)$$

We multiply both sides by  $e^{-t^2/2}$  :

$$\int_{-\infty}^t e^{-\frac{(t-x)^2}{2}} \frac{1}{\sqrt{2\pi}} g(x) dx = \frac{1}{2} \int_{-\infty}^{+\infty} e^{-\frac{(t-x)^2}{2}} \frac{1}{\sqrt{2\pi}} g(x) dx. \quad (6)$$

In other terms the unknown function  $g$  satisfies

$$\int_{-\infty}^{+\infty} \text{sign}(t-x) \varphi(t-x) g(x) dx = 0 \quad (7)$$

for all  $t \in \mathbb{R}$ . A formal derivation of (7) in  $t$ , using the product rule under the integral, and with one derivative being twice a delta function, leads to the equation

$$g(t) = \int_{-\infty}^{+\infty} q(t-x) g(x) dx \quad (8)$$

a.e. in  $t$ , where  $q(y) := \frac{1}{2}|y|e^{-\frac{y^2}{2}}$  is a probability density, but instead of justifying this formal differentiation, it seems easier to start by computing the derivative of

$$h(t) := \int_{-\infty}^t e^{tx} P(dx).$$

By Lemma 2 the distribution function  $F$  of  $P$  has a continuous inverse  $F^{-1}$ . Using the quantile transform we have

$$h(t) = \int_0^1 \mathbf{1}_{\{F^{-1} \leq t\}}(u) e^{tF^{-1}(u)} du = \int_0^{F(t)} e^{tF^{-1}(u)} du = H(F(t), t)$$

with  $H(s, t) := \int_0^s e^{tF^{-1}(u)} du$  for  $s \in (0, 1)$  and  $t \in \mathbb{R}$ . Now  $H$  has continuous partial derivatives  $H_1(s, t) = e^{tF^{-1}(s)}$  and  $H_2(s, t) = \int_0^s F^{-1}(u) e^{tF^{-1}(u)} du$ , due to the

continuity of  $F^{-1}$ , and hence  $H$  is differentiable. Let  $f$  be a Lebesgue density of  $P$ . Then, at every  $t$  where  $F'(t) = f(t)$ , and hence at Lebesgue-a.e.  $t$ , the chain rule yields

$$\begin{aligned} h'(t) &= H_1(F(t), t)f(t) + H_2(F(t), t) = e^{t^2}f(t) + \int_0^{F(t)} F^{-1}(u)e^{tF^{-1}(u)} du \\ &= e^{t^2}f(t) + \int_{-\infty}^t xe^{tx}f(x) dx. \end{aligned}$$

Thus differentiating the identity (5) and observing that  $f(x) = g(x)\varphi(x)$  we obtain the following a.e.-identity

$$\frac{1}{\sqrt{2\pi}}e^{t^2/2}g(t) + \int_{-\infty}^t xe^{tx-\frac{x^2}{2}} \frac{1}{\sqrt{2\pi}}g(x) dx = \frac{1}{2} \int_{-\infty}^{+\infty} xe^{tx-\frac{x^2}{2}} \frac{1}{\sqrt{2\pi}}g(x) dx,$$

and multiplying the latter by  $\sqrt{2\pi}e^{-t^2/2}$  gives

$$g(t) = \frac{1}{2} \left( \int_t^{+\infty} xe^{-(t-x)^2/2} g(x) dx - \int_{-\infty}^t xe^{-(t-x)^2/2} g(x) dx \right).$$

Adding to the right hand side above the quantity

$$0 = \frac{t}{2} \left( \int_{-\infty}^t e^{-(t-x)^2/2} g(x) dx - \int_t^{+\infty} e^{-(t-x)^2/2} g(x) dx \right)$$

(recall (6)) yields the desired (8).

Next, with the (positive) Radon measures  $\mu(dx) := g(x)dx$  and  $\sigma(dx) := q(x)dx$ , equation (8) can be rewritten as the so-called Choquet-Deny equation  $\mu = \mu * \sigma$ . Observe that  $t \mapsto \int_{-\infty}^{+\infty} e^{tx} \sigma(dx)$  is even and strictly convex, and is therefore equal to 1 only at  $t = 0$ . We can now use the results in section 6 of Deny (1960), where “ $n > 1$ ” is evidently a misprint for “ $n \geq 1$ ”, to conclude that  $\mu$  has to be a positive scalar multiple of the Lebesgue measure. Since  $g$  is a probability density with respect to a probability measure, we have  $g = 1$  a.e., and the theorem is proved.  $\square$

Finally, it is worthwhile to mention a natural conjecture about exponential families which seems harder to establish:

**Conjecture.** Suppose that the probability  $P$  satisfies (1), and denote  $m(t) := \int_{\mathbb{R}} xP_t(dx)$ . If for all  $t$  real  $m(t)$  is a median of  $P_t$ , then  $P = N(m, \sigma^2)$  for some  $m$  and  $\sigma$ .

This conjecture, which is probably more meaningful from a methodological point of view than the result established in the paper, does not translate in a neat harmonic analysis statement as (7) and (8) and as such it seems harder to establish. The next simple result offers some support to the conjecture. A probability  $Q$  on  $\mathbb{R}^n$  is said to be symmetric if there exists some  $m \in \mathbb{R}^n$  such that  $X - m \sim m - X$  when  $X \sim Q$ .

**Theorem 2.** *Let  $P$  be a probability on  $\mathbb{R}^n$  such that*

$$L(t) = \int_{\mathbb{R}^n} e^{\langle t, x \rangle} P(dx)$$

*is finite for all  $t \in \mathbb{R}^n$ . Assume that for all  $t \in \mathbb{R}^n$  the probability  $P_t(dx) = e^{\langle t, x \rangle} P(dx)/L(t)$  is symmetric. Then  $P$  is normal.*

**Proof.** Clearly  $m(t) = \int_{\mathbb{R}^n} x P_t(dx) = L'(t)/L(t)$  exists and, since  $P_t$  is symmetric,  $X_t - m(t) \sim m(t) - X_t$  when  $X_t \sim P_t$ . Therefore its Laplace transform

$$s \mapsto \mathbb{E}(e^{\langle s, X_t - m(t) \rangle}) = e^{-\langle s, m(t) \rangle} \frac{L(t+s)}{L(t)}$$

does not change when we replace  $s$  by  $-s$ . Considering the logarithm and taking the derivative in  $s$  we get  $2m(t) = m(t+s) + m(t-s)$ . Taking again the derivative in  $s$  we get  $m'(t+s) = m'(t-s)$  for all  $t, s \in \mathbb{R}^n$ , which means that  $m'$  is constant, hence  $\log L$  is polynomial of degree at most 2, and hence  $P$  is normal.  $\square$

### 3 References

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