

UNIVERSAL ESTIMATORS OF A VECTOR PARAMETER

by

A. L. RUKHIN  
Purdue University

Department of Statistics  
Division of Mathematical Sciences  
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## ABSTRACT

Let  $\underline{x}$  be a random sample with a distribution depending on a vector parameter  $\theta \in \mathbb{R}^m$ . The description of distributions and generalized prior densities on  $\mathbb{R}^m$  is given, for which the generalized Bayes estimator of  $\theta$ , based on  $\underline{x}$ , is the same for all symmetric loss functions. These distributions form a special subclass of exponential family. The specification of this result for the case of a location parameter is considered. The proof of the main theorem is based on the solution of a functional equation of D'Alembert's type.

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A. L. Rukhin  
Purdue University

## 1. Introduction

Let  $P_\theta$ ,  $\theta \in \Theta$ , be a family of probability measures on an abstract space  $X$ , such that each distribution  $P_\theta$  is absolutely continuous with respect to some  $\sigma$ -finite measure  $\mu$  on  $X$ . We assume throughout the paper that the coincidence of distributions  $P_{\theta_1}$  and  $P_{\theta_2}$  implies  $\theta_1 = \theta_2$ , and that  $\Theta$  is an open connected subset of the Euclidean space  $\mathbb{R}^m$ . Let  $\lambda$  be a generalized prior density on  $\Theta$ , and define

$$\pi_{\underline{x}}(\theta) = \left( \prod_{j=1}^n p(x_j, \theta) \right) \lambda(\theta)$$

where  $p(x, \theta) = \frac{dP_\theta}{d\mu}(x)$ ,  $x \in X$ ,  $\underline{x} = (x_1, \dots, x_n) \in X^n$ , and  $\theta \in \Theta$ . For convenience, we let  $\pi_{\underline{x}}(\theta)$  be zero for  $\theta \notin \Theta$ . Also, let  $W(\delta, \theta) = W(\delta - \theta)$  be the loss function depending only on the difference between the estimator  $\delta$  and the true value of the parameter  $\theta$ . Thus  $W(t)$  is defined for  $t \in \Theta - \Theta$ , and we assume that for each  $w$  the set  $\{t: W(t) \leq w\}$  is convex.

The generalized Bayes estimator  $\delta(\underline{x})$  of  $\theta$  based on the random sample  $\underline{x}$  satisfies the equation

$$\int_{\Theta} W(\delta(\underline{x}) - \theta) \pi_{\underline{x}}(\theta) d\theta = \inf_{t \in \Theta} \int_{\Theta} W(t - \theta) \pi_{\underline{x}}(\theta) d\theta. \quad (1.1)$$

In general this estimator depends on the choice of the loss function  $W$ , which is rarely known exactly to the statistician. Therefore, it seems rather natural to investigate situations in which the generalized Bayes estimator is the same for every loss from a certain set of loss functions under consideration.

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This problem for the case  $m = 1$  was solved by the author (Rukhin 1978b). The situation where  $\theta$  is a real location parameter and  $\lambda(\theta)$  is constant has been considered in [5]. In this paper we treat the case of arbitrary  $m$ .

If  $\mathcal{W}$  is a set of loss functions  $W$  such that the integrals in (1.1) converge for every  $W \in \mathcal{W}$ , then the estimator  $\delta(\underline{x})$  is called universal if  $\delta(\underline{x})$  satisfies (1.1) for all  $W \in \mathcal{W}$ . Thus a universal estimator is optimal with regard to every loss function  $W$  from  $\mathcal{W}$ .

Assume that all functions  $W$  under consideration are symmetric (i.e.  $W(-t) = W(t)$ ) and differentiable. We also suppose that differentiation with respect to  $t$  in the right side of (1.1) is allowed under the integral sign, and that the relation

$$\int W'_i(t)g(t)dt = 0$$

( $W'_i(t) = \frac{\partial}{\partial t_i} W(t)$ ,  $i=1, \dots, m$ ) valid for all  $W \in \mathcal{W}$  and a continuous function  $g(t)$ , implies that  $g(t) = g(-t)$ . Any class  $\mathcal{W}$  of loss functions satisfying these conditions we call a CS (complete symmetric) set.

If  $\delta(\underline{x})$  is universal with regard to a CS set  $\mathcal{W}$ , and  $\pi_{\underline{x}}(t)$  is a continuous function of  $t$ , then

$$\int W'_i(t - \delta(\underline{x}))\pi_{\underline{x}}(t)dt = 0.$$

Hence for all  $t \in \mathbb{R}^m$

$$\pi_{\underline{x}}(\delta(\underline{x}) - t) = \pi_{\underline{x}}(\delta(\underline{x}) + t). \quad (1.2)$$

Also, if the second derivatives of  $W(t)$  exist, the matrix  $(\int \frac{\partial^2}{\partial t_i \partial t_k} W(t)\pi_{\underline{x}}(\delta(\underline{x}) + t)dt)$  is positive semidefinite. This establishes the first part of the following proposition.

**Proposition 1.1** Assume that  $\delta(\underline{x})$  is the universal estimator of  $\theta \in \Theta \subset \mathbb{R}^m$  with regard to a CS set  $\mathcal{W}$  of loss functions. If  $\pi_{\underline{x}}(t)$  is a continuous function

of  $t \in \mathbb{R}^m$ , then (1.2) holds for all  $t \in \mathbb{R}^m$ . If (1.2) is satisfied for almost all  $t$ , then  $\delta(\underline{x})$  is the generalized Bayes estimator for every symmetric convex loss function such that the integrals in (1.1) exist.

The second part of Proposition 1.1 is well known (cf. for instance Deutsch (1965) pp. 14,16). Its proof consists in noticing that if  $W(-t) = W(t)$  and (1.2) holds, then

$$\begin{aligned} & \int W(t-\theta) \pi_{\underline{x}}(\theta) d\theta - \int W(\delta(\underline{x})-\theta) \pi_{\underline{x}}(\theta) d\theta \\ &= \int W(\theta+t') \pi_{\underline{x}}(\theta+\delta(\underline{x})) d\theta - \int W(\theta) \pi_{\underline{x}}(\theta+\delta(\underline{x})) d\theta \\ &= \frac{1}{2} \int [W(\theta+t') + W(\theta-t') - 2W(\theta)] \pi_{\underline{x}}(\theta+\delta(\underline{x})) d\theta, \end{aligned} \quad (1.3)$$

where  $t' = t - \delta(\underline{x})$ . Thus for convex functions  $W$ , the quantities in (1.3) are nonnegative, which implies (1.1).

In this paper we describe under mild regularity restrictions, the densities  $p(x, \theta)$  and generalized priors  $\lambda(\theta)$  for which the relation (1.2) holds. For these families a universal estimator exists for any CS set of loss functions.

Section 2 contains the major part of the proof of the main theorem, Theorem 2.1. In section 3 we discuss the case of a multivariate location parameter and the statistical properties of the distributions obtained. The solution of a functional equation of D'Alembert's type is needed to complete the proof of Theorem 2.1, and is given in section 4.

## 2. THE MAIN RESULT.

**THEOREM 2.1** Let  $\{p(x, \theta), x \in X, \theta \in \Theta\}$  be a family of probability densities given on a differentiable manifold  $X$  of dimension  $\rho$  where  $\Theta$  is an open symmetric connected subset of  $\mathbb{R}^m$ . Assume that, for each  $\theta \in \Theta$ ,  $p(x, \theta)$  is a positive differentiable function of  $x$ . Assume also that  $\rho \geq m$ ,

and that  $p(x, \theta)$  is continuous in  $\theta$  for each fixed  $x$ , and that  $\lambda(\theta)$  is a positive continuous function of  $\theta$ . Suppose further that for some  $n \geq 3$  and all  $\underline{x} \in X^n$  (except for  $x$  from some nowhere dense set  $N$ ) the function  $\pi_{\underline{x}}(\theta) = \prod_{j=1}^n p(x_j, \theta) \lambda(\theta)$  is symmetric with respect to the point  $\delta(\underline{x})$ , where  $\delta$  is a continuous function on  $X^n \setminus N$  and  $\delta(X^n \setminus N) = \Theta$ . Then there exist non-negative integers  $q_1, \dots, q_R$  satisfying  $q_1 + \dots + q_R = m$  such that

$$\begin{aligned} \log p(x, \theta) + \frac{1}{n} \log \lambda(\theta) &= \sum_{r=1}^R \sum_{i \leq i, k \leq q_r} d_{ik}^{(r)}(x) \\ &\times \left[ e^{\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} P_i^{(r)} \left( \frac{t-b(x)}{2} \right) - e^{-\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} P_i^{(r)} \left( \frac{b(x)-t}{2} \right) \right] \\ &\times \left[ e^{\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} P_k^{(r)} \left( \frac{t-b(x)}{2} \right) - e^{-\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} P_k^{(r)} \left( \frac{t-b(x)}{2} \right) \right]. \end{aligned}$$

Here the  $P_k^{(r)}(-t) = -P_k^{(r)}(t)$ , are polynomials with complex coefficients of degree  $p_k^{(r)}$ , and  $p_k^{(1)} \leq 2q_1 - 1$ ,  $k=1, \dots, q_1$ ;  $p_k^{(r)} \leq q_r - 1$ ,  $j=1, \dots, q_r$ ,  $r=2, \dots, R$ ;  $\alpha_1=0$ ,  $\alpha_r \in \mathbb{C}$ ,  $\alpha_r=0$ ,  $r=2, \dots, R$ ; and  $d_{ik}^{(r)}(x) = d_{ki}^{(r)}(x)$ ,  $1 \leq i, k \leq q_r$ ,  $r=1, \dots, R$ , and  $b(x)$  are real functions of  $x$ . The coefficients  $r F_{j_1 \dots j_m}^{(k)}$  of the polynomial  $P_k^{(r)}$  satisfy the relations (2.7) and (2.8).

Proof. For  $x \in X$  and  $s \in \Theta$  define

$$\psi(x, s) = \log p(x, s) + \frac{1}{n} \log \lambda(s).$$

Then from the assumption that  $\underline{x} \notin N$ , it follows that

$$\sum_{j=1}^n \psi(x_j, \delta(\underline{x}) - s) = \sum_{j=1}^n \psi(x_j, \delta(\underline{x}) + s).$$

Let

$$w_{\sigma}(t) = (2\pi)^{-m/2} \prod_{i=1}^m \sigma_i^{-1} \exp\left\{-\frac{1}{2} \sum_{i=1}^m \frac{t_i^2}{\sigma_i^2}\right\},$$

and

$$\psi_{\sigma}(x, s) = \int \psi(x, s-t) w_{\sigma}(t) dt.$$

Then  $\psi_\sigma(x, s)$  is infinitely differentiable in  $s$  and  $\psi_\sigma(x, s) \rightarrow \psi(x, s)$  as  $\sigma \rightarrow 0$ .

Observe that for  $\underline{x} \notin N$

$$\begin{aligned}
 \sum_1^n \psi_\sigma(x_j, \delta(\underline{x}) - s) &= \sum_1^n \int \psi(x_j, \delta(\underline{x}) - s - t) w_\sigma(t) dt \\
 &= \sum_1^n \int \psi(x_j, \delta(\underline{x}) + s + t) w_\sigma(t) dt \\
 &= \sum_1^n \int \psi(x_j, \delta(\underline{x}) + s - t) w_\sigma(t) dt \\
 &= \sum_1^n \psi_\sigma(x_j, \delta(\underline{x}) + s).
 \end{aligned} \tag{2.1}$$

For fixed  $\underline{x} \notin N$ , define

$$\bar{\psi}_\sigma(s) = \sum_{j=1}^n \psi_\sigma(x_j, \delta(\underline{x}) + s).$$

We first claim that there exist non-zero numbers  $\tau_1, \dots, \tau_m$  such that

$$D(\sigma, \tau) = \det \left\{ \frac{1}{2\tau_i} \frac{\partial}{\partial s_k} [\bar{\psi}_\sigma(s + \tau_i e_i) - \bar{\psi}_\sigma(s - \tau_i e_i)] \Big|_{s=0} \right\} = 0,$$

where  $e_1, \dots, e_m$  is a basis of  $\mathbb{R}^m$  and  $\tau = (\tau_1, \dots, \tau_m)$ . Indeed  $D(\sigma, \tau)$  is an analytic function of  $\sigma$  for  $\sigma \neq 0$ , and the limit of  $D(\sigma, \tau)$  as

$\tau \rightarrow 0$  is  $\det \left[ \frac{\partial^2}{\partial s_i \partial s_k} \bar{\psi}_\sigma(s) \Big|_{s=0} \right]$ . This determinant does not vanish for small  $\sigma$ , since otherwise there exists an  $i, i=1, \dots, m$ , such that

$$\int \left[ -1 + \frac{s_i^2}{\sigma_i^2} \right] \exp \left\{ -\frac{1}{2} \sum_1^m \frac{s_i^2}{\sigma_i^2} \right\} \sum_1^n \psi(x_j, \delta(\underline{x}) + s) ds = 0$$

for all  $\sigma$ . By known properties of the Laplace transform, this relation implies that  $\frac{\partial}{\partial s_i} \sum_1^n \psi(x_j, \delta(\underline{x}) + s) = 0$ , which is impossible. Thus there exist linearly independent vectors  $t_i = \tau_i e_i, i=1, \dots, m$ , such that for all sufficiently small  $\sigma$

$$\det\left\{\frac{\partial}{\partial s_k} \sum_{j=1}^n [\psi_\sigma(x_j, s+t_i) - \psi_\sigma(x_j, s-t_i)]\right\}_{s=\delta(\underline{x})} \neq 0. \quad (2.2)$$

Denote  $\phi(x, t) = (\psi_\sigma(x, t+t_1) - \psi_\sigma(x, t-t_1), \dots, \psi_\sigma(x, t+t_m) - \psi_\sigma(x, t-t_m))'$ .

(The symbol ' denotes transposition). It follows from (2.2) that, for  $\underline{x} \in N$ ,  $\sum_{j=1}^n \psi(x_j, s)$  is a local isomorphism at the point  $s = \delta(\underline{x})$ . In other words, there exists a neighborhood  $V$  of  $\delta(\underline{x})$  such that the restriction of  $\sum_{j=1}^n \psi(x_j, s)$  to  $V$  establishes an isomorphism between  $V$  and an open subset of  $\mathbb{R}^m$ . Thus the relation

$$\sum_{j=1}^n \psi_\sigma(x_j, t+t_i) = \sum_{j=1}^n \psi_\sigma(x_j, t-t_i) \quad (2.3)$$

for  $i=1, \dots, m$  and  $t \in V$  implies that  $t = \delta(\underline{x})$ .

Note also that if  $x = (x^1, \dots, x^\rho)$ ,  $\rho \geq m$ , where the  $x^i$  are coordinates of  $x$ , then the matrix  $(\frac{\partial}{\partial x^i} \phi(x, t))$  has rank  $m$  for all  $t$ . (If this rank were less than  $m$  for some  $t$ , an application of the rank theorem (cf. Narasimhan (1968)) shows that there exist two diffeomorphisms  $g_1$  and  $g_2$  such that  $g_1 \cdot \phi \cdot g_2$  has the form  $(x^1, \dots, x^r, 0, \dots, 0)$  where  $r < m$ . Then  $\sum_{j=1}^n \phi(x_j, t)$  could not be a local isomorphism.)

Let  $\Delta_t = \{\underline{x} : \underline{x} \in X/N, \delta(\underline{x}) = t\}$ , so that  $X/N = \bigcup_t \Delta_t$ . Also let  $X_t$  be the projection of  $\Delta_t$  on  $X$ ,  $X_t = \{x : \exists x_2, \dots, x_n, \delta(x, x_2, \dots, x_n) = t\}$ . Then  $X \setminus \bigcup_t X_t$  is nowhere dense. We next show that the set  $T_x = \{t : x \in X_t\}$  contains a nonempty open set if  $T_x$  is nonempty.

For every  $s \in T_x$ , there exists, because of (2.3), a neighborhood  $\mathcal{V}_s$  of  $s$  such that  $\mathcal{V}_s \cap T_x = \{t : t \in \mathcal{V}_s, \phi(x, t) + \phi(x_2, t) + \dots + \phi(x_n, t) = 0$  for some  $x_2, \dots, x_n\}$ . The implicit function theorem and the proven fact concerning the rank of the matrix  $(\frac{\partial}{\partial x^i} \phi(x, t))$  imply that  $\mathcal{V}_s \cap T_x$  contains a neighborhood of  $s$ . Since  $T_x$  is nonempty except for  $x$  in a nowhere dense set, it follows that  $T_x$  contains a nonempty open set except for  $x$  in a nowhere dense set.

For fixed  $t$  and  $x \in X_t$ , the relation  $\phi(x,t) = \phi(x_1,t)$  implies that  $\psi_\sigma(x,t+s) - \psi_\sigma(x,t-s) = \psi_\sigma(x_1,t+s) - \psi_\sigma(x_1,t-s)$  for all  $s$ . In fact, for some  $x_2, \dots, x_n$

$$\phi(x,t) = - \sum_2^n \phi(x_j,t) = \phi(x_1,t)$$

and  $\delta(x_1, x_2, \dots, x_n) = t = \delta(x, x_2, \dots, x_n)$ . Therefore

$$\psi_\sigma(x,t+s) - \psi_\sigma(x,t-s) = - \sum_2^n [\psi_\sigma(x_j,t+s) - \psi_\sigma(x_j,t-s)] = \psi_\sigma(x_1,t+s) - \psi_\sigma(x_1,t-s).$$

Thus for every  $s$  and  $t$  there exists a real function  $g$  such that

$$\psi_\sigma(x,t+s) - \psi_\sigma(x,t-s) = g(\phi(x,t)).$$

It can be easily proven that  $g$  is continuous, and that

$$g(z_1) + \dots + g(z_n) = 0$$

if  $z_1 + \dots + z_n = 0$ . Since  $n \geq 3$  it follows from this equation that

$$g(z_1 + z_2) = g(z_1) + g(z_2).$$

Because of the continuity of  $g$  there exists a vector  $h \in \mathbb{R}^m$  such that  $g(z) = \langle h, z \rangle$  (cf. Aczel (1966) p.302). Thus we have proved that

$$\psi_\sigma(x,t+s) - \psi_\sigma(x,t-s) = \sum_{i=1}^m h_i(t,s) \phi_i(x,t). \quad (2.4)$$

Note that (2.4) is true for all  $s$  and  $x \in X_t$ , where the set  $T_x = \{t: x \in X_t\}$  contains a neighborhood except for  $x$  from a nowhere dense set.

It follows from (2.1) that for those  $x_j \in X_t$  satisfying  $\delta(x_1, \dots, x_n) = t_0$

$$\sum_1^n \psi_\sigma(x_j,t) = \sum_1^n \psi_\sigma(x_j, 2t_0 - t).$$

Hence

$$\begin{aligned} \sum_1^m h_i(t,s) \sum_1^n \phi_i(x_j,t) &= \sum_1^n [\psi_\sigma(x_j,t+s) - \psi_\sigma(x_j,t-s)] \\ &= \sum_1^n [\psi_\sigma(x_j, 2t_0 - t - s) - \psi_\sigma(x_j, 2t_0 - t + s)] \end{aligned}$$

$$\begin{aligned}
&= -\sum_1^m h_i(2t_0-t, s) \sum_1^n \phi_i(x_j, 2t_0-t) \\
&= \sum_1^m h_i(2t_0-t, s) \sum_1^n \phi_i(x_j, t).
\end{aligned}$$

The last relation follows from the equalities

$$\begin{aligned}
\sum_1^n \phi_i(x_j, 2t_0-t) &= \sum_1^n [\psi_\sigma(x_j, 2t_0-t+t_i) - \psi_\sigma(x_j, 2t_0-t-t_i)] \\
&= -\sum_1^n [\psi_\sigma(x_j, t-t_i) - \psi_\sigma(x_j, t+t_i)] = -\sum_1^n \phi_i(x_j, t).
\end{aligned}$$

Thus

$$\sum_1^m [h_i(t, s) - h_i(2t_0-t, s)] \sum_1^n \phi_i(x_j, t) = 0.$$

This relation implies that  $h_i(t, s) = h_i(2t_0-t, s)$  for all  $i=1, \dots, m$ . In fact since the rank of the matrix  $(\frac{\partial}{\partial x_i} \phi(x, t))$  is equal to  $m$ , one can find  $\underline{x}^{(1)}, \dots, \underline{x}^{(m)} \in X^n/N$  such that the vectors  $\sum_1^n \phi(\underline{x}_j^{(k)}, t)$  are linearly independent. Therefore  $h_i(t, s) = h_i(s)$ ,  $i=1, \dots, m$ , and the functional equation (2.4) can be rewritten in the following form:

$$\psi_\sigma(x, t+s) - \psi_\sigma(x, t-s) = \sum_1^m h_i(s) \phi_i(x, t).$$

It is clear that if  $\phi(x, t) = 0$ , then  $\psi_\sigma(x, t+s) = \psi_\sigma(x, t-s)$  for all  $s$ .

Let  $\varphi_\sigma(x, t) = \psi_\sigma(x, t+b(x))$ , where  $\phi(x, b(x)) = 0$ . Then  $\varphi(x, -t) = \varphi(x, t)$  and

$$\varphi(x, t+s) - \varphi(x, t-s) = \sum_{i=1}^m h_i(s) \phi_i(x, t). \quad (2.5)$$

Now it follows from the proof of Theorem 4.1 that there exist nonnegative integers  $q_1, \dots, q_R$  ( $q_1 + \dots + q_R = m$ ), such that

$$\begin{aligned}
\varphi(x, t) &= \sum_{r=1}^R \sum_{1 \leq i, k \leq q_r} d_{ik}^{(r)}(x) \\
&\times [e^{\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} p_i^{(r)}(\frac{t-b(x)}{2}) - e^{-\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} p_i^{(r)}(\frac{b(x)-t}{2})] \\
&\times [e^{\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} p_k^{(r)}(\frac{t-b(x)}{2}) - e^{-\frac{1}{2} \langle \alpha_r, t-b(x) \rangle} p_k^{(r)}(\frac{b(x)-t}{2})], \quad (2.6)
\end{aligned}$$

where  $p_k^{(r)}(t) = \sum_r F_{i_1 \dots i_m} \frac{t_1^{i_1} \dots t_m^{i_m}}{i_1! \dots i_m!}$ ,  $p_k^{(r)}(-t) = -p_k^{(r)}(t)$  are polynomials with complex coefficients of degree  $p_k^{(r)}$ ,  $p_k^{(1)} \leq 2q_1 - 1$ ;  $p_k^{(r)} \leq q_r - 1$ ,  $r \geq 2$ ,  $\alpha_1 = 0$ ,  $\alpha_r \neq 0$ ,  $r \geq 2$ ,  $\alpha_r \in \mathbb{C}^m$ ;  $d_{ik}^{(r)}(x) = d_{ki}^{(r)}(x)$ ,  $i, k = 1, \dots, q_r$ ,  $r = 1, \dots, R$ . Independence of the polynomials  $p_k^{(r)}$  and vectors  $\alpha_r$  of  $x$  can be seen by substituting (2.6) in (2.5). Also, there exist complex numbers  $B_{k_1 \dots k_m}^{(ki)}$ , such that

$$\sum B_{k_1 \dots k_m}^{(ki)} {}_1 F_{i_1 \dots i_m}^{(i)} = {}_1 F_{i_1 + k_1 \dots i_m + k_m}^{(k)}, \quad (2.7)$$

and for all  $x$

$$\sum_{\ell} d_{i\ell}^{(1)} B_{k_1 \dots k_m}^{(\ell k)} = \sum_{\ell} d_{k\ell}^{(1)}(x) B_{k_1 \dots k_m}^{(\ell i)}$$

or  $r \geq 2$ ,

$${}_r F_{i_1 \dots i_m}^{(\ell)} = \langle {}_r N_{i_1}^k \dots {}_r N_{i_m}^m f_r, e_{\ell} \rangle, \quad (2.8)$$

where  ${}_r N_i$  is a lower triangular matrix with zero diagonal,  $f_r$  is a fixed vector,  $e_{\ell}$  are linearly independent vectors, and the matrices  ${}_r N_i$  satisfy the relation (4.20). Letting  $\sigma$  go to zero completes the proof.

3. DISCUSSION. THE LOCATION PARAMETER CASE. It follows from Theorem 2.1 that any family of densities with a universal estimator has the form

$$p(x, \theta) = C(\theta) \exp\{\varphi(x, \theta)\},$$

where  $C(\theta) = [\lambda(\theta)]^{1/n}$  and  $\varphi(x, t)$  is given by the formula (2.6). Thus  $p(x, \theta)$  belongs to the exponential family. When the prior density  $\tilde{\lambda}(\theta)$  is from the conjugate family, i.e., has the form

$$\tilde{\lambda}(\theta) = [C(\theta)]^{-n} \exp\{\varphi(x_0, \theta)\} \quad x_0 \in X,$$

the posterior density is symmetric.

The most interesting example of the densities (3.1) arises when  $\theta$  is a location parameter. Then  $\Theta = \mathbb{R}^m = X$ ,  $\mu$  is Lebesgue measure, and

$$p(x, \theta) = p(x - \theta) = C(\theta) \exp\{\varphi(x - \theta)\},$$

so that  $C(\theta) = \text{constant}$ .

Theorem 2.1 implies that  $\varphi$  is symmetric with respect to some point  $\theta_0$ . By shifting the initial density we can assume that  $\theta = 0$ . Also if  $\varphi(c-t) = \varphi(c+t)$  for all  $t$  and some  $c$ , then  $\varphi(t-c) = \varphi(t+c)$  for all  $t$ , implying that  $c = 0$ .

From the proof of the Theorem 4.1 it is clear that

$$\varphi(x-t) = \langle D(x)h(\frac{t-b(x)}{2}), h(\frac{t-b(x)}{2}) \rangle,$$

where the matrix  $D(x)$  has elements  $d_{ik}^{(r)}(x)$ , and the vector function  $h$  arises from the polynomials  $P_k^{(r)}(x)$ . Thus for all  $t$

$$\varphi(x-b(x)-t) = \varphi(x-b(x)+t),$$

so that  $b(x) = x$ , and

$$\varphi(t) = \langle D(x)h(t/2), h(t/2) \rangle.$$

It follows that  $D(x) = D$  and all functions  $d_{ik}^{(r)}(x)$  are constant. Note that  $p(x-\theta)$  can be written

$$p(x-\theta) = C \exp\left\{ \sum_{\ell=1}^L T_{\ell}(x) Q_{\ell}(\theta) \right\},$$

with linearly independent functions  $\{T_{\ell}(x)\}$  and  $\{Q_{\ell}(\theta)\}$ , where the  $Q_{\ell}(\theta)$  are polynomials of the form  $\theta_1^{\ell_1} \dots \theta_m^{\ell_m}$ ,  $\ell_1 + \dots + \ell_m \leq \max_{r,j} p_j^{(r)}$ , or have one of the two forms  $e^{\langle \alpha_r, \theta \rangle} \theta_1^{\ell_1} \dots \theta_m^{\ell_m}$ ,  $\ell_1 + \dots + \ell_m \leq \max_j p_j^{(r)}$  or  $e^{-\langle \alpha_r, \theta \rangle} \theta_1^{\ell_1} \dots \theta_m^{\ell_m}$ ,

$\ell_1 + \dots + \ell_m < \max_j p_j^{(r)}$ . Therefore

$$L \leq \binom{2 \max_{r,j} p_j^{(r)} + m}{m} + 2 \sum_{r=2}^R \binom{2 \max_j p_j^{(r)} + m}{m}.$$

Another formula for the density  $p$  is

$$p(x) = C \exp\{ \langle \exp[-x_1 L_1 - \dots - x_m L_m] a, b \rangle \}.$$

Here  $L_1, \dots, L_m$  are commuting matrices of order  $L$ ,  $a, b$  fixed vectors. To see this, note that because of (3.2), the linear space  $\mathcal{L}$  spanned by all

functions  $\log p(\cdot - \theta)$ ,  $\theta \in \mathbb{R}^m$ , is finite-dimensional with functions  $T_1(x), \dots, T_k(x)$  constituting its basis. Define the operator  $M(\theta)$ ,  $\theta \in \mathbb{R}^m$ , acting in  $\mathcal{L}$  by  $M(\theta)\psi(\cdot) = \psi(\cdot - \theta)$ ,  $\psi \in \mathcal{L}$ . Then  $M(\theta)$  is a finite-dimensional operator, and  $M(\theta_1 + \theta_2) = M(\theta_1)M(\theta_2)$ . It follows that

$$M(\theta) = \exp\{-\theta_1 L_1 - \dots - \theta_m L_m\},$$

where the matrices  $L_1, \dots, L_m$  are commuting and have the order  $L$ . Therefore

$$\log p(x) = \langle M(x)a, b \rangle$$

for some vectors  $a, b \in \mathcal{L}$ .

Because of (2.9) the linear subspace  $\mathcal{L}_0$  of  $\mathcal{L}$  generated by the vectors  $[\exp\{t_1 L_1 + \dots + t_m L_m\} - \exp\{-t_1 L_1 - \dots - t_m L_m\}]a$ ,  $t_1, \dots, t_m \in \mathbb{R}^1$ , is of dimension  $m$ . In other words, there are exactly  $m$  linearly independent vectors among  $L_1^{k_1} \dots L_m^{k_m} a$ ,  $k_1 + \dots + k_m$  odd. Note that the subspace  $\mathcal{L}_0$ , which is spanned by latter vectors, is invariant for all operators  $L_1^2, \dots, L_m^2$ .

The symmetry condition  $\varphi(-x) = \varphi(x)$  implies that

$$\langle [M(t) - M(-t)]a, b \rangle = 0,$$

i.e. the vector  $b$  belongs to the orthogonal complement of the space  $\mathcal{L}_0$ . The universal estimator  $\delta(\underline{x})$  satisfies the relation

$$\langle [M(t) - M(-t)]a, \sum_1^n M^*(x_j - \delta(\underline{x}))b \rangle = 0.$$

#### 4. THE FUNCTIONAL EQUATION OF D'ALEMBERT'S TYPE

The theorem proved in this section gives a solution of the following functional equation:

$$\varphi(s+t) - \varphi(s-t) = \sum_{i=1}^m h_i(s) k_i(t). \quad (4.1)$$

Here  $\varphi$  is a real continuous function defined on an open connected subset  $\mathcal{D}$  (containing zero) of Euclidean space  $\mathbb{R}^k$  and (4.1) holds for some continuous

functions  $h_i, k_i, i=1, \dots, m$ , and for all  $s, t, s+t, s-t \in \mathcal{D}$ . Without loss of generality it can be assumed that  $h_i(s)$  and  $k_i(t)$  are linearly independent. (Otherwise we can replace in (4.1) the linearly dependent functions with their representations as linear combinations relative to a basis of linearly independent functions, the result of which is an equation of similar form).

Let  $\langle \alpha, \beta \rangle = \sum_{i=1}^m \alpha_i \bar{\beta}_i$  denote the inner product of any two vectors  $\alpha, \beta \in \mathbb{C}^m$ . With this notation, the equation (4.1) can be rewritten as

$$\varphi(s+t) - \varphi(s-t) = \langle h(s), k(t) \rangle,$$

where  $h'(s) = (h_1(s), \dots, h_m(s))$  and  $k'(t) = (k_1(t), \dots, k_m(t))$ .

**Theorem 4.1** Let  $\varphi$  be an even continuous function defined on a symmetric open connected subset of  $\mathbb{R}^k$ ,  $\theta \in \mathcal{D}$ , which satisfies (4.1) for some linearly independent functions  $h_i$  and  $k_i, i=1, \dots, m$ . Then there exist nonnegative integers  $q_1, \dots, q_R, q_1 + \dots + q_R = m$ , such that  $\varphi$  admits the following representation:

$$\varphi(s) = \sum_{j=1}^{q_1} d_j^{(1)} [P_j^{(1)}(s/2)]^2 + \sum_{r=2}^R \sum_{j=1}^{q_r} d_j^{(r)} [e^{\frac{1}{2}\langle \alpha_r, s \rangle} P_j^{(r)}(s/2) - e^{-\frac{1}{2}\langle \alpha_r, s \rangle} P_j^{(r)}(-s/2)]^2,$$

where the  $P_j^{(r)}, P_j^{(r)}(-t) = -P_j^{(r)}(t)$  are polynomials with complex coefficients of degree  $p_j^{(r)}, p_j^{(1)} \leq 2q_1 - 1, j=1, \dots, q_1; p_j^{(r)} \leq q_r - 1, j=1, \dots, q_r, r=2, \dots, R; d_j^{(r)} \in \mathbb{R}^1, j=1, \dots, q_r, r=1, \dots, R; \alpha_r \in \mathbb{C}^m, \alpha_r \neq 0, r=2, \dots, R$ . The coefficients  $F_{i_1 \dots i_m}^{(j)}$  of the polynomial  $P_j^{(1)}$  satisfy the relations (4.11) for some constants  $B_{k_1 \dots k_m}^{(kj)}$  satisfying condition (4.18). The coefficients  $\phi_{i_1 \dots i_m}^{(j)}$  of the polynomial  $P_j^{(r)}, j=1, \dots, q_r, r=2, \dots, R$ , satisfy the relation (4.19) with matrices  $N_j^{(r)}$  satisfying condition (4.20). Every function  $\varphi$  of such a form is a solution of (4.1) for some functions  $k_i, h_i, i=1, \dots, m$ .

Proof. The symmetry assumption  $\varphi(-t) = \varphi(t)$  implies that  $k(-t) = -k(t)$  and that

$$\langle h(s), k(t) \rangle = \langle h(t), k(s) \rangle.$$

Note that elements  $\bar{t}_i$ ,  $i=1, \dots, m$ , can be found such that the vectors  $h(\bar{t}_i)$  are linearly independent (cf. for instance Aczel (1966) p. 201). If the matrix  $A$  is defined by the relation  $k(\bar{t}_j) = Ah(\bar{t}_j)$ , then  $k(s) = A'h(s)$  and

$$\langle Ah(s), h(t) \rangle = \langle Ah(t), h(s) \rangle.$$

Hence  $A'=A$ . Since the functions  $h_i(t)$  are linearly independent,  $A$  is non-singular, and

$$\varphi(s+t) - \varphi(s-t) = \langle Ah(s), h(t) \rangle.$$

Now for all  $s, t, u$

$$\varphi(s+t+u) - \varphi(s+t-u) = \langle Ah(s+t), h(u) \rangle$$

and

$$\varphi(s-t+u) - \varphi(s-t-u) = \langle Ah(s-t), h(u) \rangle.$$

It follows from these relations that

$$\begin{aligned} \langle A[h(s+t)+h(s-t)]h(u) \rangle &= \varphi(s+t+u) - \varphi(s-t-u) + \varphi(s-t+u) - \varphi(s+t-u) \\ &= \langle Ah(s), h(t+u) \rangle + \langle Ah(s), h(-t+u) \rangle. \end{aligned}$$

Define matrices  $A(t)$  by the formula

$$h(\bar{t}_j+t) + h(\bar{t}_j-t) = 2A(t)h(\bar{t}_j), \quad j=1, \dots, m.$$

Then

$$A[h(t+u) + h(-t+u)] = 2A'(t)Ah(u)$$

and

$$\langle Ah(s), A(t)h(u) \rangle = \langle h(s), A'(t)Ah(u) \rangle.$$

Thus for all  $t$

$$AA(t) = A'(t)A \tag{4.3}$$

and

$$h(t+u) + h(-t+u) = 2A(t)h(u). \quad (4.4)$$

Note that  $A(-t) = A(t)$  for all  $t$ . It is also clear that

$$\begin{aligned} 2A(s)A(t)h(u) &= A(s)[h(t+u)+h(-t+u)] \\ &= \frac{1}{2}[h(s+t+u)+h(-s+t+u)+h(s-t+u)+h(-s-t+u)] \\ &= [A(s+t)+A(s-t)]h(u). \end{aligned}$$

Because of this relation the matrices  $A(t)$  satisfy D'Alembert's functional equation

$$A(s+t) + A(s-t) = 2A(s)A(t). \quad (4.5)$$

An immediate consequence of (4.5) is that all matrices  $A(t)$  are commutative. It is known (cf. Suprunenko and Tyshkevich (1968) p. 16) that the whole space  $\mathbb{R}^m$  can be represented as a direct sum of invariant subspaces  $Q_r$ , with respect to all  $A(t)$ , for  $r=1, \dots, R$ . The irreducible parts of the restrictions  $A(t)|_{Q_r}$  are equivalent, while for  $r \neq w$  the irreducible parts of  $A(t)|_{Q_r}$  and  $A(t)|_{Q_w}$  are not equivalent. Because of Shur's lemma (cf. for instance [7] pp. 4,8) all irreducible parts of  $A(t)|_{Q_r}$ ,  $r=1, \dots, R$ , are one-dimensional operators. Hence all matrices  $A(t)$  can be simultaneously reduced to the form  $A(t) = T^{-1}B(t)T$ , with complex matrices  $T$  and  $B(t)$ , the latter being of the form

$$B(t) = \begin{pmatrix} B_1(t) & 0 & \dots & 0 \\ 0 & B_2(t) & \dots & 0 \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & & B_R(t) \end{pmatrix}$$

Here  $B_r(t)$ ,  $r=1, \dots, R$ , is a lower triangular matrix of dimension  $q_r = \dim Q_r$  given by

$$B_r(t) = \begin{pmatrix} b^{(r)}(t) & 0 & \dots & 0 \\ b_{21}^{(r)}(t) & b^{(r)}(t) & \dots & 0 \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ b_{q_r 1}^{(r)}(t) & b_{q_r 2}^{(r)}(t) & \dots & b^{(r)}(t) \end{pmatrix},$$

where  $b^{(r)}(t) \neq b^{(w)}(t)$ ,  $r \neq w$ , and  $q_1 + \dots + q_R = m$ .

If  $\tilde{A} = (T^{-1})^* A T^{-1}$ , then  $\tilde{A}^* = \tilde{A}$  and

$$\tilde{A}B(t) = B^*(t)\tilde{A}. \quad (4.6)$$

Therefore  $\tilde{A}$  has the form

$$\tilde{A} = \begin{pmatrix} \tilde{A}_1 & 0 & \dots & 0 \\ 0 & \tilde{A}_2 & \dots & 0 \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \dots & \tilde{A}_R \end{pmatrix},$$

with symmetric  $q_r \times q_r$  matrices  $\tilde{A}_r$ ,  $r=1, \dots, R$ . It follows from (4.4) that if  $f(s) = Th(s)$ , then

$$2B(t)f(s) = f(s+t) + f(s-t) \quad (4.7)$$

and  $\langle Ah(s), h(t) \rangle = \langle Af(s), f(t) \rangle$ . Clearly  $f(-s) = -f(s)$ .

Let  $f'(s) = (f_1(s), \dots, f_R(s))$ , where the vector-function  $f_r$  has dimension  $q_r$ ,  $r=1, \dots, R$ . The relations (4.5) and (4.7) imply that

$$B_r(s+t) + B_r(s-t) = 2B_r(s)B_r(t) \quad (4.8)$$

and

$$2B_r(t)f_r(s) = f_r(s+t) + f_r(s-t), \quad r=1, \dots, R. \quad (4.9)$$

It follows from (4.8) that

$$b^{(r)}(s+t) + b^{(r)}(s-t) = 2b^{(r)}(s)b^{(r)}(t).$$

All solutions of this D'Alembert's functional equation are known (cf. Kannappan (1968)) to be of the form

$$b^{(r)}(s) = \cosh(\alpha_1^{(r)}s_1 + \dots + \alpha_m^{(r)}s_m) = \cosh \langle \alpha_r, s \rangle, \quad (4.10)$$

for some complex numbers  $\alpha_1^{(r)}, \dots, \alpha_m^{(r)}$ .

Consider first the case of vanishing  $\alpha$ , say,  $\alpha_1=0$ , i.e.  $b^{(1)}(s)=1$ . Then  $B_1(t) = I+N(t)$ , where  $N^q(t)=0$ ,  $q=q_1$ , and  $I$  is the identity matrix. It follows from (4.8) and (4.9) that

$$N(s+t) - 2N(s) + N(s-t) = 2N(t) + 2N(t)N(s)$$

and

$$f_1(s+t) - 2f_1(s) + f_1(s-t) = 2N(t)f_1(s). \quad (4.11)$$

If one defines the operator  $L(t)$  by the relation  $L(t)f(\cdot) = f(\cdot - t)$ ,  $t \in \mathbb{R}^m$ , then (4.11) can be rewritten as

$$[L(t/2) - L(-t/2)]^2 f_1(s) = 2N(t)f_1(s).$$

It follows by induction that

$$[L(t/2) - L(-t/2)]^{2q} f_1(s) = [2N(t)]^q f_1(s) = 0,$$

so that (cf. for example Aczel (1966) p. 130) each coordinate function  $f_j^{(1)}(s)$  of  $f_1(s)$  is a polynomial. It is easy to see that  $f_j^{(1)}(s)$  is a polynomial of degree less than  $2q_1$  (Actually,  $f_j^{(1)}(s)$  is a polynomial of degree less than  $2j$ ,  $j=1, \dots, q$ .) Thus

$$f_j^{(1)}(s) = \sum_{i_\ell < 2q-1} F^{(j)}_{i_1 \dots i_m} \frac{s_1^{i_1} \dots s_m^{i_m}}{i_1! \dots i_m!},$$

where  $s'=(s_1, \dots, s_m)$ . Since  $f_1(-s)=-f_1(s)$ , we have  $F_{i_1, \dots, i_m}^{(j)}=0$  if  $\sum_{l=1}^q i_l$  is an even number. Analogously,

$$b_{kj}^{(1)}(s) = \sum_{k_\ell < 2q-2} B_{k_1 \dots k_m}^{(kj)} \frac{s_1^{k_1} \dots s_m^{k_m}}{k_1! \dots k_m!}, \quad k > j,$$

and  $B_{k_1 \dots k_m}^{(kj)}=0$  if  $\sum_{l=1}^q k_l$  is an odd number or  $\sum_{l=1}^q k_l = 0$ .

Substituting these expressions for  $f_1(s)$  and  $N(t)$  into (4.10) shows that

$$\sum_j B_{k_1 \dots k_m}^{(kj)} F_{i_1 \dots i_m}^{(j)} = F_{i_1+k_1 \dots i_m+k_m}^{(k)}. \quad (4.12)$$

If  $\tilde{A}_1 = \{\alpha_{ik}, 1 \leq i, k \leq q\}$ ,  $\alpha_{ik} = \bar{\alpha}_{ki}$ , then (4.6) implies

$$\sum_\ell \alpha_{i\ell} B_{k_1 \dots k_m}^{(\ell k)} = \sum_\ell \alpha_{k\ell} B_{k_1 \dots k_m}^{(\ell i)}. \quad (4.13)$$

Note from (4.12) and (4.13) that if  $k_j \geq \ell_j$ ,  $j=1, \dots, m$ , then

$$\sum_{i,k} \alpha_{ik} F_{i_1 \dots i_m}^{(i)} F_{k_1 \dots k_m}^{(k)} = \sum_{i,k} \alpha_{ik} F_{i_1+\ell_1 \dots i_m+\ell_m}^{(i)} F_{k_1-\ell_1 \dots k_m-\ell_m}^{(k)}$$

Hence if  $\sum_{j=1}^m (i_j+k_j) > 2m$ , it follows that

$$\sum_{i,k} \alpha_{ik} F_{i_1 \dots i_m}^{(i)} F_{k_1 \dots k_m}^{(k)} = 0. \quad (4.14)$$

Indeed we can take  $\sum_{j=1}^m i_j \geq 2m+1$ , which will imply that  $F_{i_1 \dots i_m}^{(i)} = 0$ .

Now let us return to (4.10) and consider the case of non-zero  $\alpha$ , say,  $\alpha_2 = \alpha \neq 0$ ,  $q_2 = p$ ,  $B_2(t) = B(t)$ . In this case,  $B^2(t) - I$  is nonsingular for all  $t$  such that  $\langle \alpha, t \rangle \neq 2\pi i a$  for integer  $a$ . Therefore, there exists a  $t_0$  and a nonsingular complex lower triangular matrix  $G$  such that  $G^2 = B^2(t_0) - I = \frac{1}{2}[B(2t_0) - I]$ . Indeed

$$B^2(t) - I = 2[\sinh \langle \alpha, \frac{t}{2} \rangle]^2 [I + M(t)],$$

where  $M^p(t) = 0$ . Thus we can take

$$G = \sqrt{2} \sin h \langle \alpha, \frac{t_0}{2} \rangle \left[ I + \frac{1}{2} M(t_0) + \sum_{k=2}^{p-1} \frac{(-1)^{k+1} (2k-3)!!}{2^k k!} M^k(t_0) \right].$$

Clearly  $G^2 = B^2(t_0) - I$ , and  $G$  commutes with all matrices  $B(t)$ .

Now let

$$\begin{aligned} G(t) &= G^{-1} [B(t)(G - B(t_0)) + B(t+t_0)] \\ &= B(t) - G^{-1} [B(t)B(t_0) - B(t+t_0)]. \end{aligned}$$

It is easy to check (cf. Kannapan (1968)) that

$$G(s+t) = G(s)G(t),$$

so that  $G(s) = \exp\{s_1 G_1 + \dots + s_m G_m\}$ . Here the  $G_i$  are complex triangular matrices with diagonal elements all equal to  $\alpha_i^{(2)}$ ,  $i=1, \dots, m$ .

It follows from the definition of  $G(t)$  that

$$G(t) + G(-t) = 2B(t) - G^{-1} [2B(t)B(t_0) - B(t+t_0) - B(-t+t_0)] = 2B(t).$$

Since all matrices  $B(t)$  commute, the matrices  $G_i$ ,  $i=1, \dots, m$ , commute as well.

From (4.6), one concludes that

$$\tilde{A}_2 G_i^2 = [G_i^*]^{-2} \tilde{A}_2.$$

The equation (4.7) implies that

$$f_2(s+t) + f_2(s-t) = 2 \cosh \left( \sum_{i=1}^m t_i G_i \right) f_2(s).$$

Also

$$f_2(s+t) + f_2(t-s) = 2 \cosh \left( \sum_{i=1}^m s_i G_i \right) f_2(t),$$

so that

$$f_2(s+t) = \cosh \left( \sum_{i=1}^m t_i G_i \right) f_2(s) + \cosh \left( \sum_{i=1}^m s_i G_i \right) f_2(t).$$

On the other hand, if  $\langle \alpha, s+t \rangle \neq i \frac{2a+1}{2} \pi$ ,  $a$  an integer, then

$$\begin{aligned} f_2(s) + f_2(t) &= 2 \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i - t_i) G_i \right) f_2 \left( \frac{s+t}{2} \right) \\ &= \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i - t_i) G_i \right) \left[ \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) \right]^{-1} f_2(s+t). \end{aligned}$$

Therefore

$$\begin{aligned} & \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i - t_i) G_i \right) \left[ \cosh \left( \sum_{i=1}^m t_i G_i \right) f_2(s) + \cosh \left( \sum_{i=1}^m s_i G_i \right) f_2(t) \right] \\ &= \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) [f_2(s) + f_2(t)], \end{aligned}$$

or

$$\begin{aligned} & \left[ \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) + \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i - 3t_i) G_i \right) \right] f_2(s) \\ &+ \left[ \cosh \left( \frac{1}{2} \sum_{i=1}^m (3s_i - t_i) G_i \right) + \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) \right] f_2(t) \\ &= 2 \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) [f_2(s) + f_2(t)]. \end{aligned}$$

Combining these relations one obtains

$$\begin{aligned} & \left[ \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) - \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i - 3t_i) G_i \right) \right] f_2(s) \\ &+ \left[ \cosh \left( \frac{1}{2} \sum_{i=1}^m (s_i + t_i) G_i \right) - \cosh \left( \frac{1}{2} \sum_{i=1}^m (3s_i - t_i) G_i \right) \right] f_2(t) = 0 \end{aligned}$$

or

$$\sinh \left( \frac{1}{2} \sum_{i=1}^m (s_i - t_i) G_i \right) \sinh \left( \sum_{i=1}^m t_i G_i \right) f_2(s) = \sinh \left( \sum_{i=1}^m s_i G_i \right) \sinh \left( \frac{1}{2} \sum_{i=1}^m (s_i - t_i) G_i \right) f_2(t).$$

This implies that

$$f_2(t) = \sinh \left( \sum_{i=1}^m t_i G_i \right) f,$$

where  $f$  is a constant vector.

Since  $G_i = \alpha_i I + N_i$ , where  $N_i = N_i^{(2)}$  and  $N_i^p = 0$ ,  $i=1, \dots, m$ ,  $\exp\left\{\sum_{i=1}^m t_i G_i\right\} = \exp\left\{\sum_{i=1}^m t_i \alpha_i\right\} \sum \frac{t_1^{k_1} \dots t_m^{k_m}}{k_1! \dots k_m!} N_1^{k_1} \dots N_m^{k_m}$ . Clearly  $N_1^{k_1} \dots N_m^{k_m} = 0$  if  $\sum_{i=1}^m k_i \geq p$ .

Therefore

$$f_j^{(2)}(t) = \frac{1}{2} \left[ \exp \left\{ \sum_{i=1}^m \alpha_i t_i \right\} P_j^{(2)}(t) - \exp \left\{ - \sum_{i=1}^m \alpha_i t_i \right\} P_j^{(2)}(-t) \right], \quad j=1, \dots, p,$$

where  $P_j^{(2)}(t)$  is a polynomial of degree less than  $p$ ,

$$P_j^{(2)}(t) = \sum_{\sum k_i \leq p-1} \frac{t_1^{k_1} \dots t_m^{k_m}}{k_1! \dots k_m!} \langle N_1^{k_1} \dots N_m^{k_m} f, e_j \rangle = \sum_{\sum k_i \leq p-1} \frac{t_1^{k_1} \dots t_m^{k_m}}{k_1! \dots k_m!} \phi_{k_1 \dots k_m}^{(j)} \quad (4.15)$$

the  $e_j$  are basis vectors, and

$$2\alpha_i \tilde{A}_2 N_i + \tilde{A}_2 N_i^2 = 2\alpha_i N_i^* \tilde{A}_2 + [N_i^*]^2 \tilde{A}_2, \quad i=1, \dots, m. \quad (4.16)$$

We can, finally, give the formula for  $\varphi(s)$ . Because of (4.2),

$$\begin{aligned} \varphi(2s) &= \langle Ah(s), h(s) \rangle = \langle \tilde{A}f(s), f(s) \rangle = \sum_{1 \leq i, j \leq q_1} \alpha_{ij}^{(1)} P_i^{(1)}(s) P_j^{(1)}(s) \\ &+ \sum_{r=2}^R \sum_{1 \leq i, j \leq q_r} \alpha_{ij}^{(r)} [e^{\alpha_1^{(r)} s_1 + \dots + \alpha_m^{(r)} s_m} s_{\mathbb{N}P}^{(i)}(s) e^{-\alpha_1^{(r)} s_1 - \dots - \alpha_m^{(r)} s_m} \times P^{(i)}(-s)] \\ &\times [e^{\alpha_1^{(r)} s_1 + \dots + \alpha_m^{(r)} s_m} s_{\mathbb{N}P}^{(j)}(s) e^{-\alpha_1^{(r)} s_1 - \dots - \alpha_m^{(r)} s_m} \times P^{(j)}(-s)]. \end{aligned} \quad (4.17)$$

Here the coefficients  $F_{i_1 \dots i_m}^{(j)}$  of the polynomial  $P_j^{(r)}$  satisfy (4.12) for some quantities  $B_{k_1 \dots k_m}^{(kj)}$  which satisfy condition (4.13) (with  $\alpha_{i\ell}$  replaced by  $\alpha_{i\ell}^{(1)}$ ).

The coefficients  $\phi_{k_1 \dots k_m}^{(j)}$  are defined by (4.15) where the matrices  $N_i$  satisfy (4.16) (with general index  $r$  instead of 2). Note that because of (4.14) the degree of the first term in (4.17) does not exceed  $2q_1$ .

The formula for  $\varphi(s)$ , given in the theorem, follows from (4.17) since  $\tilde{A}_r = U_r D_r U_r^*$  with unitary matrix  $U_r$  and diagonal matrix  $D_r$ ,  $r=1, \dots, R$ . In this case (4.13) must be changed to

$$d_i^{(1)} B_{k_1 \dots k_m}^{(ik)} = d_k^{(1)} B_{k_1 \dots k_m}^{(ki)}, \quad (4.18)$$

and (4.15), (4.16) to

$$\begin{aligned} P_j^{(r)}(t) &= \sum_{\sum k_i \leq q_r - 1} \frac{t_1^{k_1} \dots t_m^{k_m}}{k_1! \dots k_m!} \langle [N_1^{(r)}]^{k_1} \dots [N_m^{(r)}]^{k_m} f_r, e_j^{(r)} \rangle \\ &= \sum_{\sum k_i \leq q_r - 1} \frac{t_1^{k_1} \dots t_m^{k_m}}{k_1! \dots k_m!} r \phi_{k_1 \dots k_m}^{(j)}, \end{aligned} \quad (4.19)$$

where  $[N_i^{(r)}]^{q_r} = 0$ ,  $i=1, \dots, m$ ,  $r=2, \dots, R$ , and

$$2\alpha_i^{(r)} D_r N_i^{(r)} + D_r [N_i^{(r)}]^2 = 2\alpha_i^{(r)} N_i^{(r)*} D_r + [N_i^{(r)*}]^2 D_r, \quad (4.20)$$

where  $D_r$  is the diagonal matrix with elements  $\{d_j^{(r)}, j=1, \dots, q_r\}$ .

It remains only to prove that every function  $\varphi$  of the form (4.17) satisfies (4.1). From our argument it follows that the matrix  $B(t)$  is nonsingular on an everywhere dense set. For such  $t$  it follows from (4.7) that

$$2f(t) = B^{-1}(t)f(2t),$$

and that

$$f(s) + f(t) = 2B\left(\frac{s-t}{2}\right)f\left(\frac{s+t}{2}\right) = B\left(\frac{s-t}{2}\right)B^{-1}\left(\frac{s+t}{2}\right)f(s+t)$$

if  $B\left(\frac{s+t}{2}\right)$  is nonsingular.

It is clear that if

$$\varphi(2s) = \langle \tilde{A}f(s), f(s) \rangle,$$

where  $\tilde{A}$  satisfies (4.6), then

$$\varphi(2s) - \varphi(2t) = \langle \tilde{A}f(s+t), f(s-t) \rangle. \quad (4.21)$$

Indeed for  $s$  and  $t$  such that  $B\left(\frac{s+t}{2}\right)$  and  $B\left(\frac{s-t}{2}\right)$  are nonsingular matrices

$$\begin{aligned} & \langle \tilde{A}f(s+t), f(s-t) \rangle \\ &= \langle \tilde{A}B\left(\frac{s+t}{2}\right)B^{-1}\left(\frac{s-t}{2}\right)[f(s)+f(t)], B\left(\frac{s-t}{2}\right)B^{-1}\left(\frac{s+t}{2}\right)[f(s)-f(t)] \rangle \\ &= \langle B^{-1*}\left(\frac{s-t}{2}\right)B^*\left(\frac{s+t}{2}\right)A[f(s)+f(t)], B\left(\frac{s-t}{2}\right)B^{-1}\left(\frac{s+t}{2}\right)[f(s)-f(t)] \rangle \\ &= \langle A[f(s)+f(t)], f(s)-f(t) \rangle = \langle Af(s), f(s) \rangle - \langle Af(t), f(t) \rangle. \end{aligned}$$

Since  $f$  is continuous, the relation (4.21) is true for all  $s$  and  $t$ . This completes the proof.

Corollary. If  $m=1$ , then the given solutions of (4.1) reduce to the known ones (cf. [1] p. 175):  $\varphi(s) = a \cosh \langle \alpha, s \rangle + b$ , or  $\varphi(s) = \langle \beta, s \rangle^2 + d$ .

If  $m=2$ , all solutions of (4.1) have one of the following forms:

$$\begin{aligned} \varphi(s) = & \alpha_1 s_1^2 + \alpha_2 s_1 s_2 + \alpha_3 s_2^2, \alpha_1 \neq 0; \varphi(s) = \beta_1 s_1^2 + \beta_2 s_1 s_2 + \beta_3 s_2^2; \text{ or } \varphi(s) = \\ & \gamma_1 [\sinh(\delta_1 s_1 + \delta_2 s_2)]^2 + \gamma_2 (\xi_1 s_1 + \xi_2 s_2) \sinh(\delta_1 s_1 + \delta_2 s_2) \times \cosh(\delta_1 s_1 + \delta_2 s_2) + \\ & \gamma_3 (\xi_1 s_1 + \xi_2 s_2)^2 [\cosh(\delta_1 s_1 + \delta_2 s_2)]^2, \text{ where } \gamma_3 \delta_1 \xi_1 = 0 \text{ and } \gamma_3 \delta_2 \xi_2 = 0. \text{ This} \end{aligned}$$

shows that the solutions of (4.1) corresponding to a certain  $m$  are not necessarily linear combinations of solutions corresponding to smaller  $m$ .

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