

## 1 Notation and asymptotic distribution of $p_k$

Let  $x$  be a  $p$ -dimensional vector variable with finite higher order moments  $\nu_k = E(x - \mu)^{\otimes k}$  ( $a^{\otimes k} = a \otimes \dots \otimes a$  is a kronecker product with  $k$  terms). Note that  $\nu_1 = 0$ .

Let  $x_1, \dots, x_n$  be an iid sample of  $x$  and

$$p_k = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^{\otimes k}, \quad k \geq 1$$

By the Law of Large numbers,

$$p_k \xrightarrow{P} \nu_k, \tag{1}$$

and by the Central Limit theorem,

$$\sqrt{n}(p_k - \nu_k) \xrightarrow{D} \mathcal{N}(0, \text{cov}((x_i - \mu)^{\otimes k})), \tag{2}$$

since the  $(x_i - \mu)^{\otimes k}$  are iid with variance matrix  $\text{cov}((x_i - \mu)^{\otimes k})$ , assumed to be finite. This implies that the asymptotic variance matrix  $\text{acov}(p_k) = \text{cov}((x_i - \mu)^{\otimes k})$  can be estimated consistently by the sample variance matrix  $\text{scov}((x_i - \mu)^{\otimes k})$  of the  $(x_i - \mu)^{\otimes k}$ 's.

Let us now define  $q_1 = \bar{x}$  and

$$q_k = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^{\otimes k}, \quad k \geq 2 \tag{3}$$

We will show that like  $p_k$ ,  $q_k$  is an asymptotically normal vector with the same asymptotic limit  $\nu_k$  as  $p_k$  (when  $k \neq 1$ ) but with an asymptotic variance matrix that may differ from the one of  $p_k$  when  $k > 2$ . We also derive an expression for a consistent estimator of  $\text{acov}(q)$  when  $q$  is an stacked vector of  $q_k$ 's.

## 2 Consistent estimation of $\Gamma_p = \text{acov}(p)$

Let

$$p = \begin{pmatrix} p_1 \\ \vdots \\ p_m \end{pmatrix}$$

Note that

$$\sqrt{n} \begin{pmatrix} p_1 - \nu_1 \\ \vdots \\ p_m - \nu_m \end{pmatrix} = \frac{1}{\sqrt{n}} \sum_i \begin{pmatrix} x_i - \mu - \nu_1 \\ \vdots \\ (x_i - \mu)^{\otimes m} - \nu_m \end{pmatrix}$$

By the central limit theorem the asymptotic covariance matrix for  $p$  is

$$\Gamma_p = \text{cov} \begin{pmatrix} x - \mu \\ \vdots \\ (x - \mu)^{\otimes m} \end{pmatrix}$$

This consists of sub-matrices of the form

$$\Gamma_{rs} = \text{scov}((x - \mu)^{\otimes r}, (x - \mu)^{\otimes s}) = E((x - \mu)^{\otimes r})((x - \mu)^{\otimes s})' - E((x - \mu)^{\otimes r})E((x - \mu)^{\otimes s})'$$

It follows that

$$\text{vec}(\Gamma_{rs}) = E(x - \mu)^{\otimes(r+s)} - E((x - \mu)^{\otimes r}) \otimes E(x - \mu)^{\otimes s} = \nu_{r+s} - \nu_r \otimes \nu_s \stackrel{a}{=} q_{r+s} - q_r \otimes q_s$$

This provides a consistent estimator for  $\Gamma_{rs}$  and these provide a consistent estimator for  $\Gamma$ . With a bit of work, we can show that this consistent estimator of  $\Gamma_p$  is the sample covariance matrix  $\text{scov}(t_i^+)$ , where

$$t_i^+ = \begin{pmatrix} x_i - \bar{x} \\ (x_i - \bar{x})^{\otimes 2} \\ \vdots \\ (x_i - \bar{x})^{\otimes k} \end{pmatrix}$$

Unfortunately, what we need is not a consistent estimator of  $\Gamma_p$  but a consistent estimator of  $\Gamma_q$ . This estimator is to be developed below.

### 3 Representation of $q_k$ as a function of $p_1$ and $p_k$

Replacing in (3)  $x_i - \bar{x}$  by  $(x_i - \mu) - (\bar{x} - \mu)$ , and expanding the kronecker power<sup>1</sup> yields

$$(x_i - \bar{x})^{\otimes k} = (x_i - \mu)^{\otimes k} - \sum_{j=1}^k (x_i - \mu)^{\otimes(j-1)} \otimes (\bar{x} - \mu) \otimes (x_i - \mu)^{\otimes(k-j)} + r_{ki} \quad (4)$$

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<sup>1</sup>Note that the binomial theorem can not be used here since kronecker power is not commutative in general; e.g.,  $(a + b)^{\otimes 2} \neq a^{\otimes 2} + 2a \otimes b + b^{\otimes 2}$ , for general vectors  $a$  and  $b$

where it is understood that  $a^{\otimes 0} = 1$ . In formula (4),  $r_{ki}$  is a kronecker product of  $k$  terms which can be shown to have the term  $(x_i - \mu)$  repeated twice or more. For developments to follow, we need the commutation matrix  $K_{mn}$  (see Magnus and Neudecker, 1986) defined by

$$K_{mn} \text{vec}(A) = \text{vec}(A'),$$

for any  $m \times n$  matrix  $A$ . It is known that

$$K_{mn} = \sum_{i=1}^m \sum_{j=1}^n (H_{ij} \otimes H'_{ji})$$

where  $H_{ij}$  is an  $m \times n$  matrix with a 1 in the  $ij$ th position and zeros elsewhere. If  $a$  and  $b$  are columns vectors of length  $n$  and  $m$  then

$$K_{mn}(a \otimes b) = b \otimes a \quad (5)$$

The property (5) permits writing (4) as

$$(x_i - \bar{x})^{\otimes k} = (x_i - \mu)^{\otimes k} - C_k (x_i - \mu)^{\otimes(k-1)} \otimes (\bar{x} - \mu) + r_{ki}, \quad (6)$$

where

$$C_k = \sum_{j=1}^{k-1} K_{p^j p^{k-j}} + I_{p^k}, k \geq 2 \quad (7)$$

To achieve this result we used that for  $j = 1, \dots, k-1$

$$\begin{aligned} & (x_i - \mu)^{\otimes(j-1)} \otimes (\bar{x} - \mu) \otimes (x_i - \mu)^{\otimes(k-j)} \\ &= \left[ (x_i - \mu)^{\otimes(j-1)} \otimes (\bar{x} - \mu) \right] \otimes (x_i - \mu)^{\otimes(k-j)} \\ &= K_{p^j p^{k-j}} \left( (x_i - \mu)^{\otimes(k-j)} \otimes \left[ (x_i - \mu)^{\otimes(j-1)} \otimes (\bar{x} - \mu) \right] \right) \\ &= K_{p^j p^{k-j}} \left( (x_i - \mu)^{\otimes(k-j)} \otimes (x_i - \mu)^{\otimes(j-1)} \otimes (\bar{x} - \mu) \right) \\ &= K_{p^j p^{k-j}} \left( (x_i - \mu)^{\otimes(k-1)} \otimes (\bar{x} - \mu) \right) \end{aligned}$$

in virtue of the associative property of the Kronecker product and (5). Clearly, for  $j = k$

$$\begin{aligned} & (x_i - \mu)^{\otimes(j-1)} \otimes (\bar{x} - \mu) \otimes (x_i - \mu)^{\otimes(k-j)} \\ &= (x_i - \mu)^{\otimes(k-1)} \otimes (\bar{x} - \mu) \end{aligned}$$

Averaging (6) across  $i$  and using the definition of the  $p_k$ 's, we obtain

$$q_k = p_k - C_k (p_{k-1} \otimes p_1) + r_k \quad (8)$$

where

$$r_k = \frac{1}{n} \sum_{i=1}^n r_{ki}$$

Subtracting  $\nu_k$  in both sides of (8) and multiplying by  $\sqrt{n}$ , yields

$$\sqrt{n}(q_k - \nu_k) = \sqrt{n}(p_k - \nu_k) - C_k(p_{k-1} \otimes \sqrt{n}p_1) + \sqrt{n}r_k, k \geq 2; \quad (9)$$

thus,

$$\sqrt{n}(q_k - \nu_k) = \begin{pmatrix} -C_k(p_{k-1} \otimes I_p) & I_{p^k} \end{pmatrix} \sqrt{n} \begin{pmatrix} p_1 - \nu_1 \\ p_k - \nu_k \end{pmatrix} + \sqrt{n}r_k, k \geq 2 \quad (10)$$

since  $\nu_1 = 0$  and

$$C_k(p_{k-1} \otimes \sqrt{n}(p_1 - \nu_1)) = C_k \text{vec}(\sqrt{n}(p_1 - \nu_1)p'_{k-1}) = C_k(p_{k-1} \otimes I_p)\sqrt{n}(p_1 - \nu_1).$$

We now need proving that

$$\sqrt{n}r_k \xrightarrow{P} 0, k \geq 2; \quad (11)$$

that is,  $\sqrt{n}r_k = o_p(1)$ . This is equivalent to proving

$$\frac{1}{\sqrt{n}} \sum r_{ki} \xrightarrow{P} 0 \quad (12)$$

For this we need a little more precise definition of  $r_{ki}$ . Let  $r_{ki}$  be the sum of all the  $k$ th order Kronecker products containing terms of the form  $(x_i - \mu)$  or  $(\bar{x} - \mu)$  with at least two of these being  $(\bar{x} - \mu)$ . Each of these is a permutation of

$$(\bar{x} - \mu)^{\otimes j} \otimes (x_i - \mu)^{\otimes(k-j)}$$

for some  $j \geq 2$ . It follows that

$$\frac{1}{\sqrt{n}} \sum_i (\bar{x} - \mu)^{\otimes j} \otimes (x_i - \mu)^{\otimes(k-j)} = \sqrt{n}(\bar{x} - \mu)^{\otimes j} \otimes \frac{1}{n} \sum_i (x_i - \mu)^{\otimes(k-j)} \xrightarrow{P} 0$$

and (11) follows from this.

Combining (1), (8) and (11), we obtain

$$q_k \xrightarrow{P} \nu_k, \quad k > 1 \quad (13)$$

That is,  $p_k$  and  $q_k$  have the same asymptotic limit when  $k > 1$ ; when  $k = 1$ , then  $q_1 \xrightarrow{P} \mu$  and  $p_1 \xrightarrow{P} \nu_1 = 0$ .

By setting  $k = 2$  in (9) and using (11) and  $p_1 \xrightarrow{P} 0$ , we obtain

$$\sqrt{n}(q_2 - \nu_2) = \sqrt{n}(p_2 - \nu_2) + o_p(1). \quad (14)$$

When  $k = 1$  then, clearly,

$$\sqrt{n}(q_1 - \mu_1) = \sqrt{n}(p_1 - \nu_1) \quad (15)$$

The results (10), (14) and (15) will be exploited in the next section to derive the asymptotic distribution of a stacked vector of  $q_k$ 's.

## 4 Consistent estimation of $\Gamma_q$

For simplicity of exposition, consider  $q = (q_1, q_2, q_3, q_4)'$ , i.e. we consider the case of  $k = 4$  (larger values for  $k$  would be handled by analogy). By stacking equations (15) and (14), and (10) with  $k = 3, 4$ , and denoting  $\nu = (\nu_1, \nu_2, \nu_3, \nu_k)'$  and  $\tilde{\nu} = (\mu_1, \nu_2, \nu_3, \nu_k)'$ , since  $p \xrightarrow{P} \nu_k$  and  $q \xrightarrow{P} \tilde{\nu}_k$ , we obtain

$$\sqrt{n} \begin{pmatrix} q_1 - \mu_1 \\ q_2 - \nu_2 \\ q_3 - \nu_3 \\ q_4 - \nu_4 \end{pmatrix} = \begin{pmatrix} I_p & 0 & 0 & 0 \\ 0 & I_{p^2} & 0 & 0 \\ -C_k(p_2 \otimes I_p) & 0 & I_{p^3} & 0 \\ -C_k(p_3 \otimes I_p) & 0 & 0 & I_{p^4} \end{pmatrix} \sqrt{n} \begin{pmatrix} p_1 - \nu_1 \\ p_2 - \nu_2 \\ p_3 - \nu_3 \\ p_4 - \nu_4 \end{pmatrix} + o_p(1); \quad (16)$$

which, by direct application of Slutsky's Theorem, proves that  $\sqrt{n}(q - \tilde{\nu})$  is asymptotically normal with variance matrix

$$\Gamma_q = U\Gamma_p U', \quad (17)$$

where

$$U = \begin{pmatrix} I_p & 0 & 0 & 0 \\ 0 & I_{p^2} & 0 & 0 \\ -C_k(\nu_2 \otimes I_p) & 0 & I_{p^3} & 0 \\ -C_k(\nu_3 \otimes I_p) & 0 & 0 & I_{p^4} \end{pmatrix}$$

Since

$$\hat{U} = \begin{pmatrix} I_p & 0 & 0 & 0 \\ 0 & I_{p^2} & 0 & 0 \\ -C_k(q_2 \otimes I_p) & 0 & I_{p^3} & 0 \\ -C_k(q_3 \otimes I_p) & 0 & 0 & I_{p^4} \end{pmatrix}$$

is a consistent estimator of  $U$  (we used (13)), a consistent estimate of  $\Gamma_q$  is

$$\hat{\Gamma}_q = \hat{U}[\text{scov}(t_i^+)]\hat{U}' = \text{scov}(\hat{U}t_i^+)$$

That is,  $\hat{\Gamma}_q$  can be written as the sample variance matrix of the pseudo-values

$$\tilde{t}_i^+ = \begin{pmatrix} I_p & 0 & 0 & 0 \\ 0 & I_{p^2} & 0 & 0 \\ -C_k(q_2 \otimes I_p) & 0 & I_{p^3} & 0 \\ -C_k(q_3 \otimes I_p) & 0 & 0 & I_{p^4} \end{pmatrix} \begin{pmatrix} x_i - \bar{x} \\ (x_i - \bar{x})^{\otimes 2} \\ (x_i - \bar{x})^{\otimes 3} \\ (x_i - \bar{x})^{\otimes 4} \end{pmatrix} \quad (18)$$

With a bit of work it can be seen that  $\text{scov}(\tilde{t}_i^+)$  coincides with the accelerated version of the IJK estimator of variance proposed in our paper, the “accelerating” matrix  $C_k$  being now expressed in terms of the commutation matrices  $K_{mn}$ . Using the  $K_{mn}$ 's can be less efficient (computationally) than using the matrix  $C_k$  defined in the Appendix of the paper.