

ORTHOGONAL SERIES ESTIMATION OF THE PAIR CORRELATION FUNCTION OF A SPATIAL POINT PROCESS

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Supplementary Material

S1 Expanding observation window

This section states a few results on the asymptotic behavior of the edge correction $e_n(h)$ (defined in (4.9) in the main document) and related ratios.

For each $n \geq 1$ and $h, h_1, h_2 \in \mathbb{R}^d$,

$$|W_n \cap (W_n)_h| = \begin{cases} \prod_{i=1}^d (2na_i - |h_i|) & |h_i| < 2na_i, i = 1, \dots, d \\ 0 & \text{otherwise} \end{cases}$$

and

$$|W_n \cap (W_n)_{h_1} \cap (W_n)_{h_2}| = \begin{cases} V_n(h_1, h_2) & \max\{|h_{1i}|, |h_{2i}|, |h_{1i} - h_{2i}|\} < 2na_i, i = 1, \dots, d \\ 0 & \text{otherwise} \end{cases}$$

where $V_n(h_1, h_2) = \prod_{i=1}^d (2na_i - \max\{0, h_{1i}, h_{2i}\} + \min\{0, h_{1i}, h_{2i}\})$. Therefore, for any fixed $h, h_1, h_2 \in \mathbb{R}^d$, as $n \rightarrow \infty$,

$$e_n(h) = \frac{|W_n \cap (W_n)_h|}{|W_n|} = \prod_{i=1}^d \left(1 - \frac{|h_i|}{2na_i}\right) \rightarrow 1. \quad (\text{S1.1})$$

If $n \geq R/\min_{1 \leq i \leq d} a_i$ or equivalently $B_{r_{\min}}^R \subset W_n$, then $1/2^d \leq e_n(h) \leq 1$ for any $h \in B_{r_{\min}}^R$.

Further,

$$\frac{|W_n \cap (W_n)_{h_1} \cap (W_n)_{h_2}|}{|W_n|} = \prod_{i=1}^d \left(1 - \frac{\max\{0, h_{1i}, h_{2i}\} - \min\{0, h_{1i}, h_{2i}\}}{2na_i}\right) \rightarrow 1. \quad (\text{S1.2})$$

Similarly, it can be shown that for any fixed $h_1, h_2, h_3 \in \mathbb{R}^d$,

$$|W_n \cap (W_n)_{h_1} \cap (W_n)_{h_3} \cap (W_n)_{h_2+h_3}| / |W_n| \rightarrow 1.$$

S2 Proofs

S2.1 Proof of Lemma 1

For n large enough, $\mathbb{I}(h \in B_{r_{\min}}^R) > 0$ implies $|W_n \cap (W_n)_h| > 0$. Then, by the second order Campbell formula (see Møller and Waagepetersen, 2003,

Section C.2.1),

$$\begin{aligned}
 \mathbb{E}(\hat{\theta}_{k,n}) &= \frac{1}{\varsigma_d |W_n|} \int_{W_n^2} \frac{\phi_k(\|v - u\| - r_{\min}) w(\|v - u\| - r_{\min})}{\|v - u\|^{d-1} e_n(v - u)} \\
 &\quad \mathbb{I}(v - u \in B_{r_{\min}}^R) g(\|v - u\|) du dv \\
 &= \int_{\mathbb{R}^d} \frac{g(\|h\|) \phi_k(\|h\| - r_{\min}) w(\|h\| - r_{\min})}{\varsigma_d \|h\|^{d-1}} \\
 &\quad \mathbb{I}(h \in B_{r_{\min}}^R) \left(\int_{\mathbb{R}^d} \frac{\mathbb{I}\{v \in W_n \cap (W_n)_h\}}{|W_n \cap (W_n)_h|} dv \right) dh \\
 &= \int_{r_{\min}}^{r_{\min}+R} g(r) \phi_k(r - r_{\min}) w(r - r_{\min}) dr = \theta_k.
 \end{aligned}$$

Let $f_k(h) = \phi_k(\|h\| - r_{\min}) w(\|h\| - r_{\min}) / \|h\|^{d-1}$. Then, by the second to fourth order Campbell formulae (omitting the details),

$$\begin{aligned}
 \text{Var}(\hat{\theta}_{k,n}) &= \frac{1}{\varsigma_d^2 |W_n|} \left[2 \int_{B_{r_{\min}}^R} g(h) f_k^2(\|h\|) G_n^{(2)}(h) dh \right. \\
 &\quad + 4 \int_{B_{r_{\min}}^R} \int_{B_{r_{\min}}^R} g^{(3)}(h_1, h_2) f_k(\|h_1\|) f_k(\|h_2\|) G_n^{(3)}(h_1, h_2) dh_1 dh_2 \\
 &\quad \left. + \int_{B_{r_{\min}}^R} \int_{B_{r_{\min}}^R} f_k(\|h_1\|) f_k(\|h_2\|) G_n^{(4)}(h_1, h_2) dh_1 dh_2 \right], \quad (\text{S2.3})
 \end{aligned}$$

where

$$\begin{aligned}
 G_n^{(2)}(h) &= \frac{1}{e_n^2(h)} \left[\frac{1}{|W_n|} \int_{W_n} \frac{\mathbf{1}[u + h \in W_n]}{\rho(u)\rho(u+h)} du \right], \\
 G_n^{(3)}(h_1, h_2) &= \frac{1}{e_n(h_1)e_n(h_2)} \left[\frac{1}{|W_n|} \int_{W_n} \frac{\mathbf{1}[u \in (W_n)_{h_1} \cap (W_n)_{h_2}]}{\rho(u)} du \right], \\
 G_n^{(4)}(h_1, h_2) &= \frac{1}{e_n(h_1)e_n(h_2)} \int_{\mathbb{R}^2} \left[g^{(4)}(h_1, u, h_2 + u) - g(h_1)g(h_2) \right] \\
 &\quad \frac{|W_n \cap (W_n)_u \cap (W_n)_{h_1} \cap (W_n)_{h_2}|}{|W_n|} du.
 \end{aligned}$$

For $n \geq R/\min_{1 \leq i \leq d} a_i$ and any $h, h_1, h_2 \in B_{r_{\min}}^R$, by V1 and Section S1,

$$\begin{aligned} G_n^{(2)}(h) &= \frac{1}{e_n^2(h)} \left[\frac{1}{|W_n|} \int_{W_n} \frac{\mathbf{1}[u+h \in W_n]}{\rho(u)\rho(u+h)} du \right] \leq \frac{1}{\rho_{\min}^2 e_n(h)} \leq \frac{2^d}{\rho_{\min}^2}, \\ G_n^{(3)}(h_1, h_2) &= \frac{1}{e_n(h_1)e_n(h_2)} \left[\frac{1}{|W_n|} \int_{W_n} \frac{\mathbf{1}[u \in (W_n)_{h_1} \cap (W_n)_{h_2}]}{\rho(u)} du \right] \\ &\leq \frac{1}{\rho_{\min} e_n(h)} \leq \frac{2^d}{\rho_{\min}} \end{aligned}$$

and by V3,

$$\begin{aligned} |G_n^{(4)}(h_1, h_2)| &\leq \frac{1}{e_n(h_1)e_n(h_2)} \int_{\mathbb{R}^2} \left| g^{(4)}(h_1, u, h_2+u) - g(h_1)g(h_2) \right| \\ &\quad \frac{|W_n \cap (W_n)_u \cap (W_n)_{h_1} \cap (W_n)_{h_2}|}{|W_n|} du \\ &\leq \frac{1}{e_n(h_1)} \int_{\mathbb{R}^2} \left| g^{(4)}(h_1, u, h_2+u) - g(h_1)g(h_2) \right| du \\ &\leq 2^d \int_{\mathbb{R}^2} \left| g^{(4)}(h_1, u, h_2+u) - g(h_1)g(h_2) \right| du \leq 2^d C_4. \end{aligned}$$

Thus, using V2, for all $k \geq 1$ and $n \geq R/\min_{1 \leq i \leq d} a_i$,

$$\begin{aligned} \mathbb{V}\text{ar}(\hat{\theta}_{k,n}) &\leq \frac{1}{\varsigma_d^2 |W_n|} \left[\frac{2^{d+1}}{\rho_{\min}^2} \int_{B_{r_{\min}}^R} f_k^2(\|h\|)g(\|h\|) dh \right. \\ &\quad + \frac{2^{d+2}}{\rho_{\min}} C_3 \left(\int_{B_{r_{\min}}^R} |f_k(\|h\|)| dh \right)^2 \\ &\quad \left. + 2^d C_4 \left(\int_{B_{r_{\min}}^R} |f_k(\|h\|)| dh \right)^2 \right]. \end{aligned}$$

But for all $k \geq 1$,

$$\begin{aligned} \int_{B_{r_{\min}}^R} f_k^2(\|h\|)g(\|h\|)dh &= \int_{r_{\min}}^{r_{\min}+R} \phi_k^2(r - r_{\min})g(r) \frac{w^2(r - r_{\min})}{r^{d-1}} dr \\ &= \int_0^R \phi_k^2(r)g(r + r_{\min}) \frac{w^2(r)}{(r + r_{\min})^{d-1}} dr \\ &\leq C_2 \int_0^R \phi_k^2(r)w(r)dr = C_2, \end{aligned}$$

and by Hölder's inequality,

$$\begin{aligned} \int_{B_{r_{\min}}^R} |f_k(\|h\|)|dh &= \int_{B_{r_{\min}}^R} |\phi_k(\|h\| - r_{\min})| \frac{w(\|h\| - r_{\min})}{\|h\|^{d-1}} dh \\ &= \int_{r_{\min}}^{r_{\min}+R} |\phi_k(r - r_{\min})| w(r - r_{\min}) dr \\ &\leq \left(\int_0^R \phi_k^2(r)w(r)dr \right)^{1/2} \left(\int_0^R w(r)dr \right)^{1/2} \\ &= \left(\int_0^R w(r)dr \right)^{1/2} < \infty. \end{aligned}$$

Therefore

$$\mathbb{V}\text{ar}(\hat{\theta}_{k,n}) < \frac{1}{\varsigma_d^2 |W_n|} \left[\frac{2^{d+1}}{\rho_{\min}^2} C_2 + \left(\frac{2^{d+2}}{\rho_{\min}} C_3 + 2^d C_4 \right) \left(\int_0^R w(r)dr \right) \right] = \frac{C_1}{|W_n|},$$

where

$$C_1 = \frac{1}{\varsigma_d^2} \left[\frac{2^{d+1}}{\rho_{\min}^2} C_2 + \left(\frac{2^{d+2}}{\rho_{\min}} C_3 + 2^d C_4 \right) \left(\int_0^R w(r)dr \right) \right] > 0.$$

S2.2 Proof of Lemma 2

Consider a real function f on $\mathbb{R}^d \times \mathbb{R}^d$ where $f(h_1, h_2) \neq 0$ implies $|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}| > 0$. Then, referring to the set-up in Section 4 in the

main document, by the fourth order Campbell formula,

$$\begin{aligned}
& \mathbb{E} \left\{ \sum_{u,v,u',v' \in X_{W_n}}^{\neq} \frac{f(v-u, v'-u')}{\rho(u)\rho(v)\rho(u')\rho(v')|W_n \cap (W_n)_{v-u}| |W_n \cap (W_n)_{v'-u'}|} \right\} \\
&= \int_{W_n^4} \frac{f(v-u, v'-u')}{|W_n \cap (W_n)_{v-u}| |W_n \cap (W_n)_{v'-u'}|} \\
&\quad g^{(4)}(v-u, u'-u, v'-u) du dv du' dv' \\
&= \int_{(\mathbb{R}^d)^4} f(h_1, h_2) g^{(4)}(h_1, u'-u, h_2 + u' - u) \\
&\quad \frac{\mathbb{I}\{u \in W_n \cap (W_n)_{h_1}, u' \in W_n \cap (W_n)_{h_2}\}}{|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}|} du dh_1 du' dh_2 \\
&= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(h_1, h_2) \\
&\quad \left\{ \frac{\int_{W_n \cap (W_n)_{h_1}} \int_{W_n \cap (W_n)_{h_2}} g^{(4)}(h_1, u'-u, h_2 + u' - u) du du'}{|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}|} \right\} dh_1 dh_2.
\end{aligned}$$

This expectation is the sum of

$$\begin{aligned}
A &= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(h_1, h_2) \left\{ \frac{\int_{W_n \cap (W_n)_{h_1}} \int_{W_n \cap (W_n)_{h_2}} g(h_1) g(h_2) du du'}{|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}|} \right\} dh_1 dh_2 \\
&= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(h_1, h_2) g(h_1) g(h_2) dh_1 dh_2
\end{aligned}$$

and

$$\begin{aligned}
 B_n &= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(h_1, h_2) \\
 &\quad \left[\frac{\int_{W_n \cap (W_n)_{h_1}} \int_{W_n \cap (W_n)_{h_2}} \{g^{(4)}(h_1, u' - u, h_2 + u' - u) - g(h_1)g(h_2)\} du du'}{|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}|} \right] dh_1 dh_2 \\
 &= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(h_1, h_2) \left[\int_{\mathbb{R}^d} \frac{|W_n \cap (W_n)_{h_1} \cap (W_n)_{h_3} \cap (W_n)_{h_2+h_3}|}{|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}|} \right. \\
 &\quad \left. \{g^{(4)}(h_1, h_3, h_2 + h_3) - g(h_1)g(h_2)\} dh_3 \right] dh_1 dh_2.
 \end{aligned}$$

We now specialize to $f(h_1, h_2) = f_k(h_1)f_k(h_2)$, where

$$f_k(h) = \phi_k(\|h\| - r_{\min}) w(\|h\| - r_{\min}) \mathbb{I}(h \in B_{r_{\min}}^R) / (\varsigma_d \|h\|^{d-1}).$$

Then, by V3,

$$\begin{aligned}
 |B_n| &\leq \int_{B_{r_{\min}}^R} \int_{B_{r_{\min}}^R} f_k(h_1) f_k(h_2) \left[\int_{\mathbb{R}^d} \frac{|W_n \cap (W_n)_{h_1} \cap (W_n)_{h_3} \cap (W_n)_{h_2+h_3}|}{|W_n \cap (W_n)_{h_1}| |W_n \cap (W_n)_{h_2}|} \right. \\
 &\quad \left. |g^{(4)}(h_1, h_3, h_2 + h_3) - g(h_1)g(h_2)| dh_3 \right] dh_1 dh_2 \\
 &\leq C_4 \int_{B_{r_{\min}}^R} \int_{B_{r_{\min}}^R} \frac{f_k(h_1) f_k(h_2)}{|W_n \cap (W_n)_{h_1}|} dh_1 dh_2.
 \end{aligned}$$

Thus B_n tends to zero as $n \rightarrow \infty$. Regarding A , we have

$$\begin{aligned}
 A &= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f_k(h_1) f_k(h_2) g(h_1) g(h_2) dh_1 dh_2 \\
 &= \left\{ \int_{\mathbb{R}^d} f_k(h) g(h) dh \right\}^2 = \left\{ \int_{r_{\min}}^{r_{\min}+R} g(r) \phi_k(r - r_{\min}) w(r - r_{\min}) dr \right\}^2 = \theta_k^2.
 \end{aligned}$$

S2.3 Proof of Theorem 1

We verify that the mean integrated squared error of $\hat{g}_{o,n}$ tends to zero as $n \rightarrow \infty$. By (3.7), $\text{MISE}(\hat{g}_{o,n}, w) / \varsigma_d = \sum_{k=1}^{\infty} [b_k(\psi_n)^2 \text{Var}(\hat{\theta}_{k,n}) + \theta_k^2 \{b_k(\psi_n) - 1\}^2]$.

By Lemma 1 and condition S1 the right hand side is bounded by

$$BC_1 |W_n|^{-1} \sum_{k=1}^{\infty} |b_k(\psi_n)| + \max_{1 \leq k \leq m} \theta_k^2 \sum_{k=1}^m (b_k(\psi_n) - 1)^2 + (B^2 + 1) \sum_{k=m+1}^{\infty} \theta_k^2.$$

By Parseval's identity, $\sum_{k=1}^{\infty} \theta_k^2 < \infty$. The last term can thus be made arbitrarily small by choosing m large enough. It also follows that θ_k^2 tends to zero as $k \rightarrow \infty$. Hence, by S2, the middle term can be made arbitrarily small by choosing n large enough for any choice of m . Finally, the first term can be made arbitrarily small by S3 and by choosing n large enough.

S2.4 Proof of Theorem 2

Let for $t = (t_1, \dots, t_d) \in \mathbb{Z}^d$,

$$\Delta(t) = \times_{i=1}^d (s(t_i - 1/2), s(t_i + 1/2])$$

be the hyper-square with side length s and centered at st . Then, $\{\Delta(t) : t \in \mathbb{Z}^d\}$ is a partition of \mathbb{R}^d ; i.e., $\Delta(t_1) \cap \Delta(t_2) = \emptyset$ for $t_1 \neq t_2$ and $\cup_{t \in \mathbb{Z}^d} \Delta(t) = \mathbb{R}^d$, and $|\Delta_n(t) \oplus R| = (s + 2R)^d$, where

$$\Delta(t) \oplus R = \times_{i=1}^d (s(t_i - 1/2) - R, s(t_i + 1/2) + R).$$

Let $\mathcal{T}_n = \{t \in \mathbb{Z}^d : \Delta(t) \cap W_n \neq \emptyset\}$ and define

$$Y_n(t) = \sum_{u \in X \cap \Delta(t)} \sum_{\substack{v \in X \setminus \{u\} \\ v-u \in B_{r_{\min}}^R}} \frac{f_n(v-u)\mathbb{I}(u \in W_n, v \in W_n)}{\rho(u)\rho(v)e_n(v-u)}.$$

Then, since $X = \bigcup_{t \in \mathbb{Z}^d} (X \cap \Delta(t))$,

$$\begin{aligned} S_n &= \frac{1}{\varsigma_d |W_n|} \sum_{\substack{u,v \in X_{W_n} \\ v-u \in B_{r_{\min}}^R}}^{\neq} \frac{f_n(v-u)}{\rho(u)\rho(v)e_n(v-u)} \\ &= \frac{1}{\varsigma_d |W_n|} \sum_{t \in \mathbb{Z}^d} \sum_{u \in X \cap \Delta(t)} \sum_{\substack{v \in X \setminus \{u\} \\ v-u \in B_{r_{\min}}^R}} \frac{f_n(v-u)\mathbb{I}(u \in W_n, v \in W_n)}{\rho(u)\rho(v)e_n(v-u)} \\ &= \frac{1}{\varsigma_d |W_n|} \sum_{\substack{t \in \mathbb{Z}^d \\ \Delta(t) \cap W_n \neq \emptyset}} Y_n(t) = \frac{1}{\varsigma_d |W_n|} \sum_{t \in \mathcal{T}_n} Y_n(t). \end{aligned}$$

Due to V1, N4 and since $e_n(h) > 1/2^d$ for n large enough and $h \in B_{r_{\min}}^R$,

$$\mathbb{E}(|Y_n(t)|^{2+\lceil \eta \rceil}) \leq \mathbb{E} \left(\sum_{u \in X \cap \Delta(t)} \sum_{\substack{v \in X \setminus \{u\} \\ v-u \in B_{r_{\min}}^R}} \frac{L_2 2^d}{\rho_{\min}^2} \right)^{2+\lceil \eta \rceil}.$$

The moments $\mathbb{E}(|Y_n(t)|^{2+\lceil \eta \rceil})$ are thus bounded by sums of integrals involving $g^{(k)}(u_1, \dots, u_{k-1})$ times $(L_2 2^d / \rho_{\min}^2)^{2+\lceil \eta \rceil}$ for $k = 2, \dots, 2(2+\lceil \eta \rceil)$. These integrals are bounded uniformly in t and n due to assumption N2. Thus,

$$\sup_{n \geq 1} \sup_{t \in \mathcal{T}_n} \mathbb{E}(|Y_n(t)|^{2+\eta}) \leq \sup_{n \geq 1} \sup_{t \in \mathcal{T}_n} \mathbb{E}(|Y_n(t)|^{2+\lceil \eta \rceil}) < \infty$$

and hence $\{|Y_n(t)|^{2+\eta} : t \in \mathcal{T}_n, n \geq 1\}$ is a uniformly integrable family (triangular array) of random variables. Invoking finally N1 and N3 and

letting $\sigma_n^2 = \text{Var}\{\sum_{t \in \mathcal{T}_n} Y_n(t)\}$, it follows directly from Theorem 3.1 in Biscio and Waagepetersen (2016) that

$$\sigma_n^{-1} \sum_{t \in \mathcal{T}_n} [Y_n(t) - \mathbb{E}\{Y_n(t)\}] \xrightarrow{D} N(0, 1),$$

which is equivalent to Theorem 2.

S3 Order of sum of products of Bessel basis functions

In this section we consider the Fourier-Bessel basis in the case $r_{\min} = 0$. It is known (see Watson, 1995, p. 199) that as $r \rightarrow \infty$,

$$J_\nu(r) \sim \left(\frac{2}{\pi r}\right)^{1/2} \cos\left(r - \frac{\nu\pi}{2} - \frac{\pi}{4}\right),$$

which implies that

$$\alpha_{\nu,k} = (k + \frac{\nu}{2} - \frac{1}{4})\pi + O(k^{-1}), \quad \text{as } k \rightarrow \infty, \quad (\text{S3.4})$$

and $\alpha_{\nu,k} \rightarrow \infty$, as $k \rightarrow \infty$. We can argue that for large k ,

$$\begin{aligned} J_{\nu+1}(\alpha_{\nu,k}) &\approx \left(\frac{2}{\pi\alpha_{\nu,k}}\right)^{1/2} \cos\left(\alpha_{\nu,k} - \frac{\pi}{4} - \frac{(\nu+1)\pi}{2}\right) \\ &= \left(\frac{2}{\pi\alpha_{\nu,k}}\right)^{1/2} \cos\left((k-1)\pi + O(k^{-1})\right), \end{aligned}$$

and consequently

$$|J_{\nu+1}(\alpha_{\nu,k})| \approx \left(\frac{2}{\pi\alpha_{\nu,k}}\right)^{1/2}. \quad (\text{S3.5})$$

On the other hand, for $\nu \geq 0$ and $r > 0$ (Landau, 2000),

$$|J_\nu(r)| \leq \frac{c}{|r|^{1/3}},$$

where $c = 0.7857468704 \dots$, and hence

$$|\phi_k(r)| \leq \frac{cr^{-\nu-1/3}\sqrt{2}}{(R^2\alpha_{\nu,k})^{1/3}|J_{\nu+1}(\alpha_{\nu,k})|} \quad r > 0, k \geq 1.$$

Using (S3.4) and (S3.5), for large k ,

$$\frac{cr^{-\nu-1/3}\sqrt{2}}{(R^2\alpha_{\nu,k})^{1/3}|J_{\nu+1}(\alpha_{\nu,k})|} \approx c\sqrt{\pi} \frac{r^{-\nu-1/3}}{R^{2/3}} \alpha_{\nu,k}^{1/6} \approx c\sqrt{\pi} \frac{r^{-\nu-1/3}}{R^{2/3}} \left\{ \left(k + \frac{\nu}{2} - \frac{1}{4} \right) \pi \right\}^{1/6}.$$

Since $\lim_{r \rightarrow 0} J_\nu(r)r^{-\nu} = \{\Gamma(\nu + 1)2^\nu\}^{-1}$, we also obtain for large k and

$$0 < \|h\| < R,$$

$$|\phi(\|h\|)| \leq \text{const} \left(\frac{R}{\alpha_{\nu,k}} \right)^{-\nu} \frac{\sqrt{2}}{R J_{\nu+1}(\alpha_{\nu,k})} \approx \text{const} \sqrt{\pi} \frac{\alpha_{\nu,k}^{\nu+1/2}}{R^{\nu+1}} = O(k^{\nu+1/2}).$$

Thus, for fixed r and $0 < \|h\| < R$,

$$|\phi_k(r)\phi_k(\|h\|)| = O(k^{1/6+\max(1/6,\nu+1/2)}) = O(k^{1/6+\max(1/6,d/2-1/2)}).$$

By generalization of Faulhaber's formula (McGown and Parks, 2007),

$$\sum_{k=1}^{K_n} k^p = O(K_n^{p+1}), \quad p > -1.$$

Therefore,

$$\frac{1}{K_n^\omega} \sum_{k=1}^{K_n} |\phi_k(r)\phi_k(\|h\|)| = \begin{cases} O(K_n^{4/3-\omega}) & d = 1 \\ O(K_n^{d/2+2/3-\omega}) & d > 1 \end{cases}$$

for $\omega > 0$ and $0 < r, \|h\| < R$.

S4 Simulation study

The results in Figure S1 are obtained as for the simulation study in the main document but with $W = [0, 2]^2$. The estimated cut-offs \hat{K} are summarized in Table S1. Simulation mean and 95% envelopes for \hat{g}_k , \hat{g}_c and \hat{g}_o with both Fourier-Bessel and cosine basis and simple smoothing schemes are shown for $W = [0, 2]^2$ in Figure S2. Figure S3 shows histograms of orthogonal series estimates. Comments on these figures and the table are given in the main document.

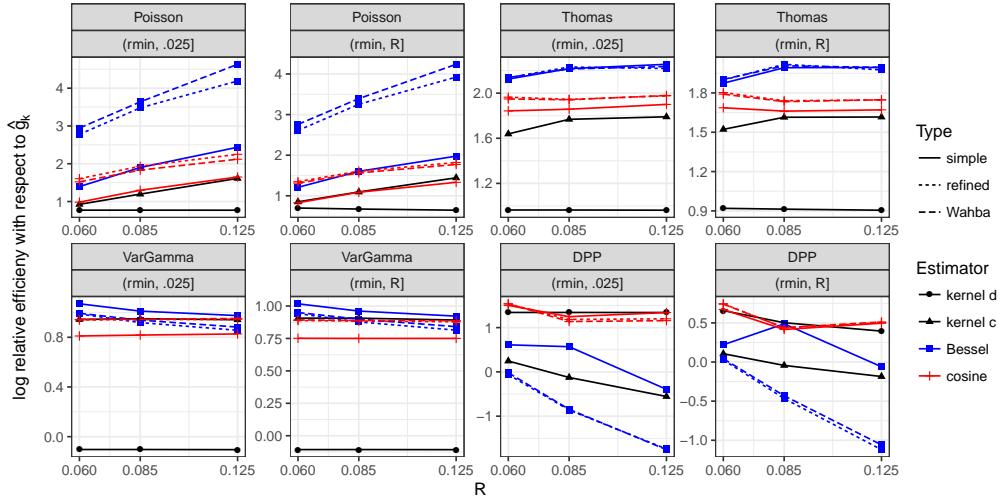


Figure S1: Plots of log relative efficiencies for small lags $(r_{\min}, 0.025]$ and all lags $(r_{\min}, R]$, $R = 0.06, 0.085, 0.125$, and $W = [0, 2]^2$. Black: kernel estimators. Blue and red: orthogonal series estimators with Bessel respectively cosine basis. Lines serve to ease visual interpretation.

S5 Data example

Figure S4 shows the data sets and fitted intensity functions considered in Section 7 of the main document. For the *Capparis frondosa* species and the orthogonal series estimator with cosine basis, the function $\hat{I}(K)$ given in (5.14) of the main document is shown in Figure S5. Although $\hat{I}(K)$ is decreasing over $1 \leq K \leq 49$, the rate of decrease slows down after $K = 7$.

S6 Behavior of the Fourier-Bessel and cosine basis

Figure S6 shows the Fourier-Bessel and cosine basis functions $\phi_k(r)$ in the planar case ($d = 2$) for $R = 0.125$, $k = 1, \dots, 8$ and $r \in [0, 0.125]$. Obviously, the cosine basis functions are uniformly bounded and integrable. However, the Fourier-Bessel basis functions exhibit damped oscillation behavior with $\phi_k(R) = 0$ and

$$\phi_k(0) = \frac{\alpha_{\nu,k}^{\nu}}{R^{\nu+1} 2^{\nu-1/2} \Gamma(\nu + 1) J_{\nu+1}(\alpha_{\nu,k})},$$

for all $k \geq 1$, because $\lim_{r \rightarrow 0} J_{\nu}(r)r^{-\nu} = 1/(\Gamma(\nu + 1)2^{\nu})$ for $\nu \geq 0$. Thus, $\phi_k(0) \rightarrow \infty$ as $k \rightarrow \infty$.

For $0 \leq \nu \leq 1/2$ (or equivalently $d = 2, 3$), $|J_{\nu}(r)| \leq (2/\pi r)^{1/2}$ and hence

$$\int_0^{\alpha_{\nu,k}} |J_{\nu}(r)| r^{\nu+1} dr \leq \left(\frac{2}{\pi}\right)^{1/2} \int_0^{\alpha_{\nu,k}} r^{\nu+\frac{1}{2}} dr = \left(\frac{2}{\pi}\right)^{1/2} \frac{(\alpha_{\nu,k})^{\nu+\frac{3}{2}}}{\nu + \frac{3}{2}}.$$

Therefore, as $k \rightarrow \infty$,

$$\begin{aligned}
\int_0^R |\phi_k(r)| w(r) dr &= \frac{\sqrt{2}}{R |J_{\nu+1}(\alpha_{\nu,k})|} \left(\frac{R}{\alpha_{\nu,k}} \right)^{\nu+2} \int_0^{\alpha_{\nu,k}} |J_\nu(r)| r^{\nu+1} dr \\
&\leq \frac{\sqrt{2}}{R |J_{\nu+1}(\alpha_{\nu,k})|} \left(\frac{R}{\alpha_{\nu,k}} \right)^{\nu+2} \left(\frac{2}{\pi} \right)^{1/2} \frac{(\alpha_{\nu,k})^{\nu+\frac{3}{2}}}{\nu + \frac{3}{2}} \\
&= \frac{2R^{\nu+1}}{|J_{\nu+1}(\alpha_{\nu,k})| (\pi \alpha_{\nu,k})^{1/2} (\nu + \frac{3}{2})} \\
&\approx \frac{2R^{\nu+1}}{(\frac{2}{\pi \alpha_{\nu,k}})^{1/2} (\pi \alpha_{\nu,k})^{1/2} (\nu + \frac{3}{2})} = \frac{\sqrt{2} R^{\nu+1}}{\nu + \frac{3}{2}} < \infty,
\end{aligned}$$

which implies uniform integrability of $\phi_k(r)$.

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Table S1: Monte Carlo mean and quantiles of the estimated cut-off \hat{K} obtained from (5.15) for the orthogonal series estimator with Fourier-Bessel (FB) and cosine (CO) bases and their ratio (CO/FB) in the case of Poisson (P), Thomas (T), Variance Gamma (V) and determinantal (D) point processes on observation windows $W_1 = [0, 1]^2$ and $W_2 = [0, 2]^2$ with $r_{\min} = 0.001$.

		$R = 0.06$			$R = 0.085$			$R = 0.125$		
		$\hat{\mathbb{E}}(\hat{K})$	$\hat{q}_{0.05}(\hat{K})$	$\hat{q}_{0.95}(\hat{K})$	$\hat{\mathbb{E}}(\hat{K})$	$\hat{q}_{0.05}(\hat{K})$	$\hat{q}_{0.95}(\hat{K})$	$\hat{\mathbb{E}}(\hat{K})$	$\hat{q}_{0.05}(\hat{K})$	$\hat{q}_{0.95}(\hat{K})$
W_1	FB	2.17	2.00	3.00	2.15	2.00	3.00	2.11	2.00	3.00
	P CO	2.31	2.00	4.00	2.35	2.00	4.00	2.34	2.00	4.00
	CO/FB	1.08	0.67	1.50	1.10	1.00	1.67	1.12	1.00	2.00
	FB	2.24	2.00	3.00	2.24	2.00	3.00	3.23	2.00	4.05
	T CO	2.39	2.00	4.00	2.52	2.00	4.00	3.54	3.00	5.00
	CO/FB	1.10	0.67	2.00	1.16	0.67	1.50	1.14	0.75	1.76
	FB	2.77	2.00	4.00	3.50	2.00	6.00	4.85	3.00	8.00
	V CO	3.06	2.00	5.00	4.17	2.00	7.00	5.78	3.00	10.00
	CO/FB	1.14	0.67	2.00	1.26	0.75	2.33	1.27	0.75	2.50
W_2	FB	2.21	2.00	3.00	2.18	2.00	3.00	2.38	2.00	3.00
	D CO	2.23	2.00	3.00	2.46	2.00	4.00	3.30	2.00	5.00
	CO/FB	1.04	0.67	1.50	1.15	1.00	1.50	1.45	1.00	2.00
	FB	2.17	2.00	3.00	2.12	2.00	3.00	2.09	2.00	3.00
	P CO	2.34	2.00	4.00	2.32	2.00	4.00	2.36	2.00	4.00
	CO/FB	1.10	0.67	2.00	1.11	0.67	2.00	1.13	1.00	2.00
	FB	2.39	2.00	4.00	2.46	2.00	4.00	3.78	3.00	5.00
	T CO	2.43	2.00	5.00	3.17	2.00	5.00	4.19	3.00	6.00
	CO/FB	1.08	0.50	2.50	1.35	1.00	2.00	1.13	0.75	1.67
V	FB	3.58	3.00	5.00	5.14	3.00	8.00	7.19	5.00	11.00
	CO	4.76	3.00	9.00	6.40	4.00	12.00	9.03	5.00	17.00
	CO/FB	1.38	0.80	2.67	1.30	0.80	2.56	1.30	0.83	2.43
D	FB	2.22	2.00	4.00	2.17	2.00	3.00	2.79	2.00	4.00
	CO	2.19	2.00	3.00	2.92	2.00	4.00	3.74	3.00	5.00
	CO/FB	1.02	0.67	1.50	1.38	1.00	2.00	1.40	1.00	2.00

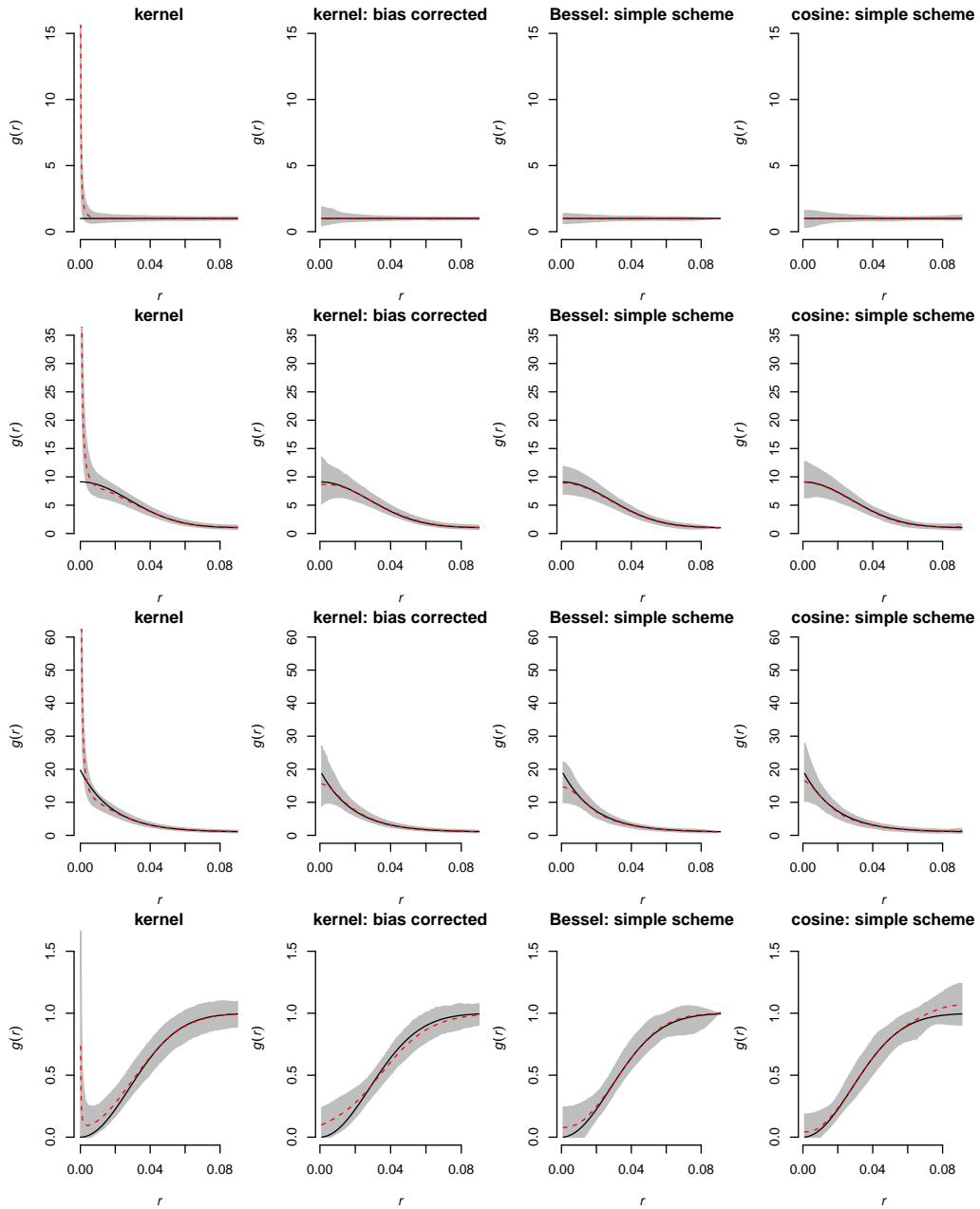


Figure S2: True pair correlation function (solid line), Monte Carlo mean (dashed lines) and 95% pointwise probability interval (grey area) of estimates based on $n_{\text{sim}} = 1000$ simulations from the Poisson (first row), Thomas (second row), Variance Gamma (third row) and determinantal (fourth row) point processes on $W = [0, 2]^2$.

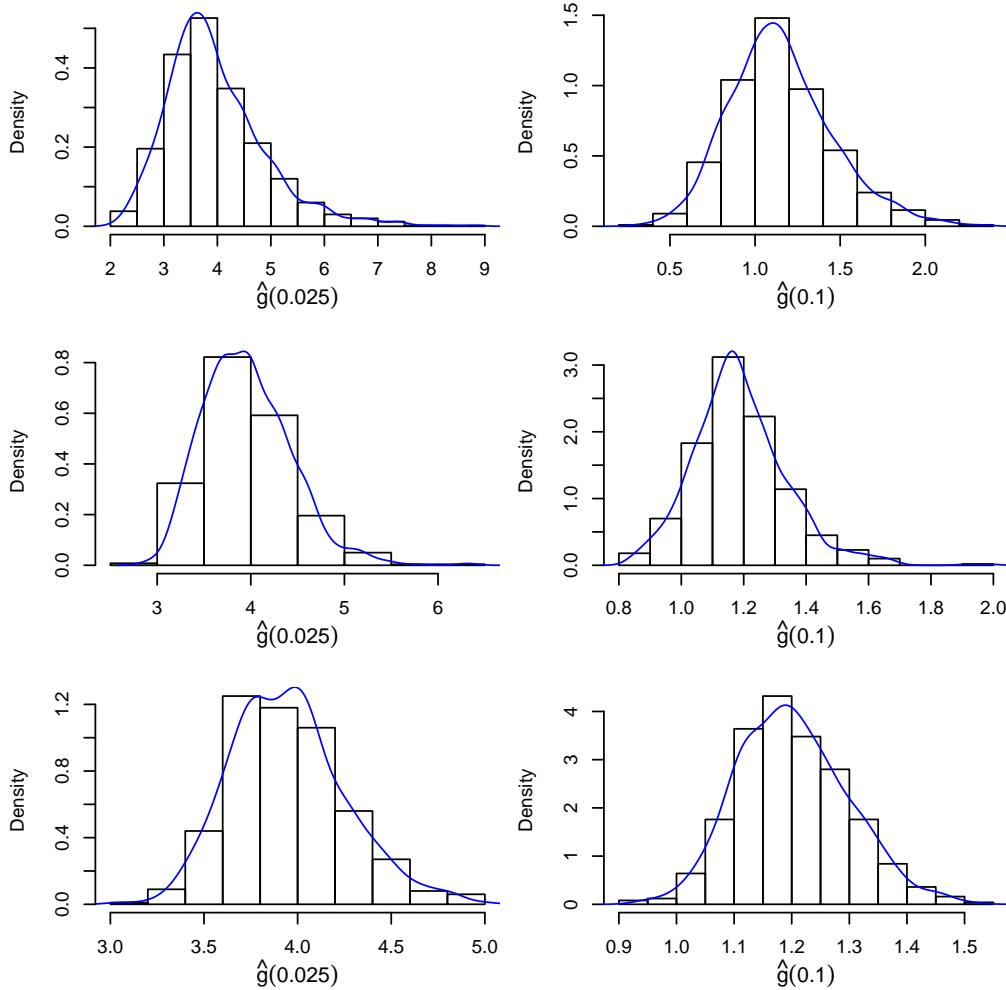


Figure S3: Histograms of $\hat{g}_o(r)$ at $r = 0.025$ and $r = 0.1$ using the Bessel basis with the simple smoothing scheme in case of the Thomas process on $W = [0, 1]^2$ (upper panels), $W = [0, 2]^2$ (middle panels) and $W = [0, 3]^2$ (lower panels).

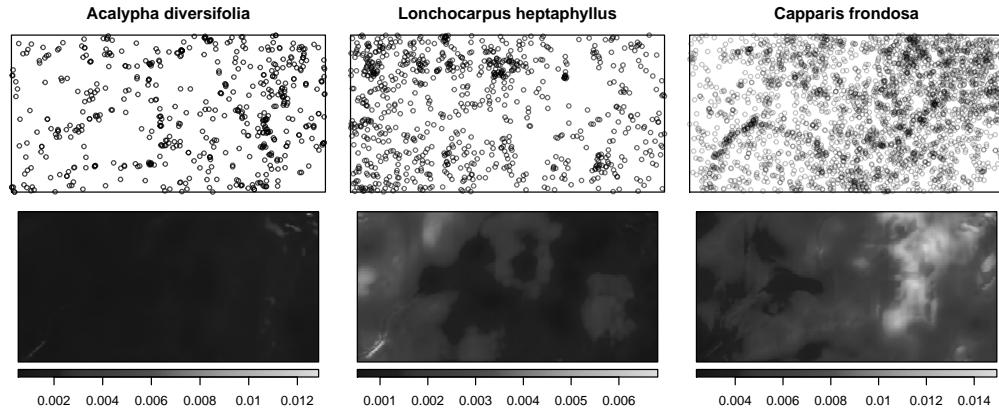


Figure S4: Locations of *Acalypha diversifolia*, *Lonchocarpus heptaphyllus* and *Capparis frondosa* trees in the Barro Colorado Island plot (upper panels) and their fitted parametric intensity functions (lower panels).

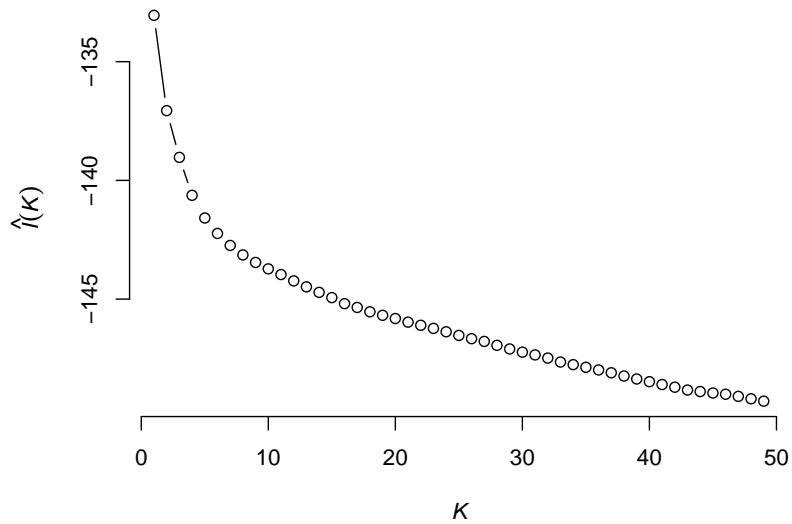


Figure S5: Estimate $\hat{I}(K)$ of the mean integrated squared error for *Capparis frondosa* in case of the orthogonal series estimator with cosine basis.

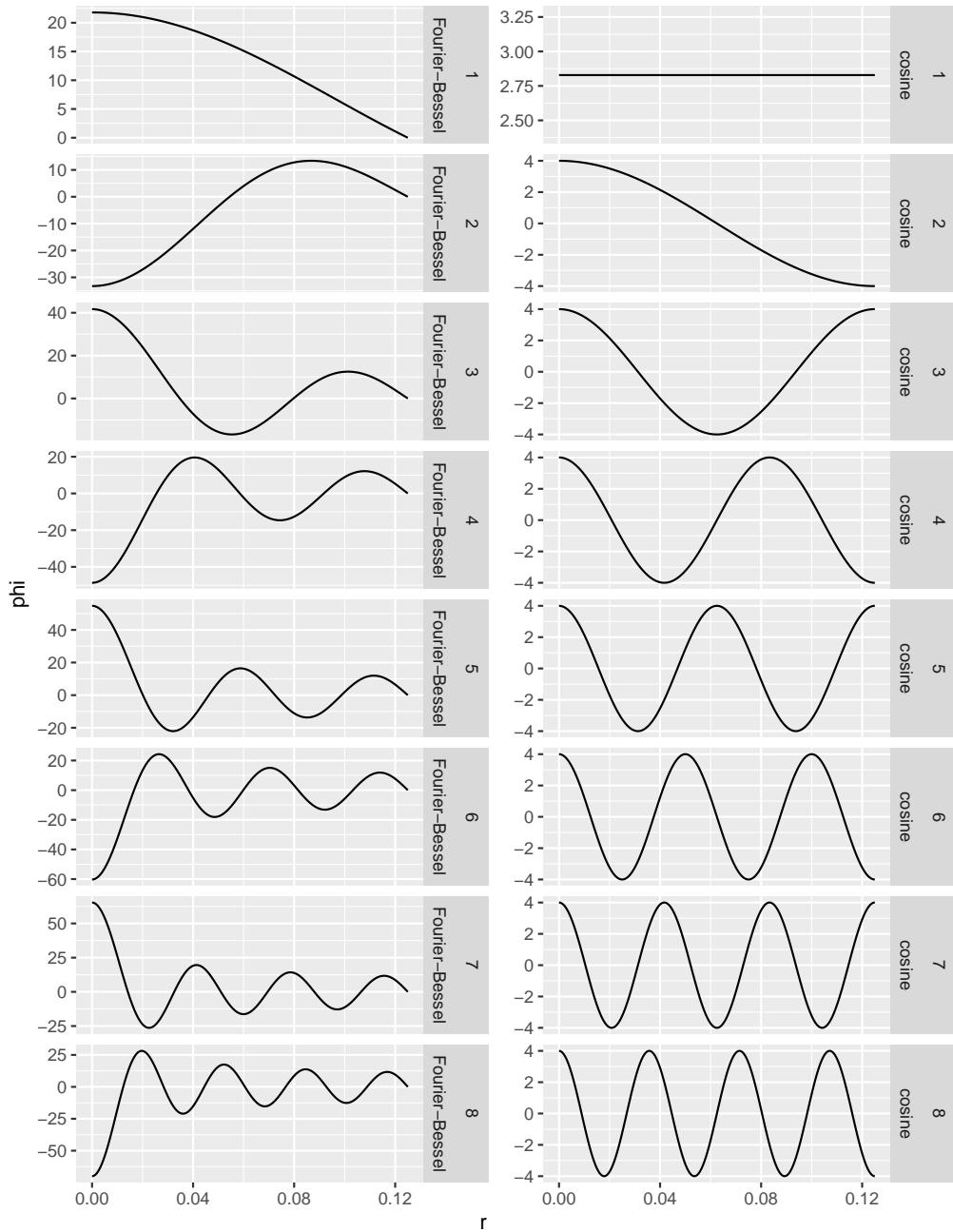


Figure S6: Fourier-Bessel and cosine basis functions $\phi_k(r)$ in the planar case ($d = 2$) for $R = 0.125$ and $k = 1, \dots, 8$.