DISCRETE SCAN STATISTICS WITH WINDOWS OF ARBITRARY SHAPE

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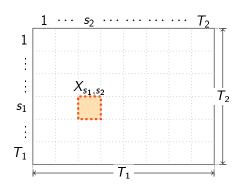
Definitions and notations





Preliminary notations

Let T_1 , T_2 be positive integers



- Rectangular region $\mathcal{R}_2 = [0, T_1] \times [0, T_2]$
- $(X_{s_1,s_2})_{1 \le s_1 \le T_1}$ i.i.d. r.v.'s $1 \le s_2 \le T_2$
 - Bernoulli($\mathcal{B}(1,p)$)
 - Binomial($\mathcal{B}(n,p)$)
 - Poisson($\mathcal{P}(\lambda)$)
 - Normal($\mathcal{N}(\mu, \sigma^2)$)
- X_{s_1,s_2} number of observed events in the elementary subregion $r_{s_1,s_2} = [s_1 - 1, s_1] \times [s_2 - 1, s_2]$





Two dimensional scan statistic

Let $2 \le m_s \le T_s$, $s \in \{1,2\}$ be positive integers

• Define for $1 \le i_s \le T_s - m_s + 1$ and $1 \le j_s \le m_s$ the 2-way tensor $\mathfrak{X}_{j_1,j_2} \in \mathbb{R}^{m_1 \times m_2}$,

$$\mathfrak{X}_{i_1,i_2}(j_1,j_2) = X_{i_1+j_1-1,i_2+j_2-1}$$

• Take $S: \mathbb{R}^{m_1 \times m_2} \to \mathbb{R}$ to be a measurable real valued function (*score function*) and define

$$Y_{i_1,i_2}(\mathcal{S}) = \mathcal{S}\left(\mathfrak{X}_{i_1,i_2}\right)$$

DEFINITION

The two dimensional scan statistic with score function ${\mathcal S}$ is defined by

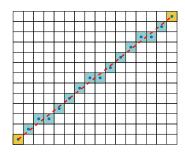
$$S_{m_1,m_2}(T_1,T_2;\mathcal{S}) = \max_{\substack{1 \leq i_1 \leq T_1 - m_1 + 1 \\ 1 \leq i_2 \leq T_2 - m_2 + 1}} Y_{i_1,i_2}(\mathcal{S})$$

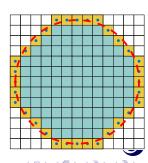


SHAPE OF THE SCANNING WINDOW

Let G be the geometrical shape of the scanning window (rectangle, quadrilateral, ellipse, etc.) and \tilde{G} be its corresponding discrete form.

- ullet Rasterization algorithms (omputer vision): continuous shape odiscrete shape
 - Line Bresenham line algorithm ([Bresenham, 1965])
 - Circle Bresenham circle algorithm ([Bresenham, 1977])
 - Bezier curves [Foley, 1995]

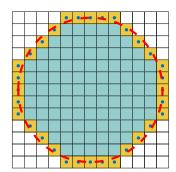




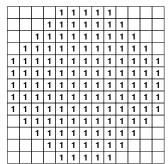
SHAPE OF THE SCANNING WINDOW

To each discrete shape \tilde{G} it corresponds an unique matrix (2-way tensor) $A(G) = A(\tilde{G})$ (of smallest size) with 0 and 1 entries (1 if there is a point and 0 otherwise):

 \tilde{G} : Circle



 $A(\tilde{G})$: Circle

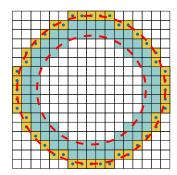




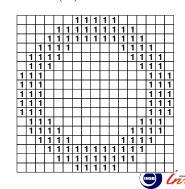
SHAPE OF THE SCANNING WINDOW

To each discrete shape \tilde{G} it corresponds an unique matrix (2-way tensor) $A\left(G
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ight)$ (of smallest size) with 0 and 1 entries (1 if there is a point and 0 otherwise):

 \tilde{G} : Annulus



 $A(\tilde{G})$: **Annulus**



Arbitrary window scan statistic

Let G be a geometric shape and A=A(G) its corresponding $\{0,1\}$ matrix of size $m_1\times m_2$.

ullet Define the score function ${\mathcal S}$ associated to the shape G by

$$S(X_{i_1,i_2}) = A \circ X_{i_1,i_2} = \sum_{s_1=i_1}^{i_1+m_1-1} \sum_{s_2=i_2}^{i_2+m_2-1} A(s_1-i_1+1,s_2-i_2+1) X_{s_1,s_2}$$

Remark

If, in particular, the shape G is a rectangle of size $m_1 \times m_2$ than its corresponding $\{0,1\}$ matrix of the same size has all the entries equal to 1 so the score function

$$S\left(\mathfrak{X}_{i_{1},i_{2}}\right) = \sum_{s_{1}=i_{1}}^{i_{1}+m_{1}-1} \sum_{s_{2}=i_{2}}^{i_{2}+m_{2}-1} X_{s_{1},s_{2}}$$

is the classical rectangular window of the two dimensional scan statistics.

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Problem and related work





OBJECTIVE

Find a good estimate for the distribution of the two dimensional discrete scan statistic with score function ${\cal S}$

$$Q_{\mathbf{m}}(\mathsf{T};\mathcal{S}) = \mathbb{P}\left(S_{\mathbf{m}}(\mathsf{T};\mathcal{S}) \leq \tau\right)$$

with $\mathbf{m}=(m_1,m_2)$ and $\mathbf{T}=(T_1,T_2)$

Previous work:

- Continuous scan statistics
 - Rectangles: [Loader, 1991], [Glaz et al., 2001], [Glaz et al., 2009]
 - Circles: [Anderson and Titterington, 1997]
 - Triangles, ellipses and other convex shapes: [Alm, 1983, Alm, 1997, Alm, 1998], [Tango and Takahashi, 2005], [Assunção et al., 2006]
- Discrete scan statistics
 - No results!



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Approximation methodology for the general scan statistic





APPROXIMATION AND ERROR BOUNDS

Theorem (Generalization of [Amărioarei, 2014])

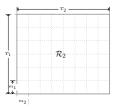
Let
$$t_1, t_2 \in \{2, 3\}$$
, $Q_{t_1, t_2} = \mathbb{P}\left(S_{\mathsf{m}}\left(t_1(m_1 - 1), t_2(m_2 - 1); \mathcal{S}\right) \leq \tau\right)$ and $L_s = \left\lfloor \frac{T_s}{m_s - 1} \right\rfloor$, $s \in \{1, 2\}$. If \hat{Q}_{t_1, t_2} is an estimate of Q_{t_1, t_2} , $\left|\hat{Q}_{t_1, t_2} - Q_{t_1, t_2}\right| \leq \beta_{t_1, t_2}$ and τ is such that $1 - \hat{Q}_{2, 2}(\tau) \leq 0.1$ then

$$\left| \mathbb{P}(S_{\mathsf{m}}(\mathsf{T}; \mathcal{S}) \leq \tau) - \left(2\hat{Q}_2 - \hat{Q}_3 \right) \left[1 + \hat{Q}_2 - \hat{Q}_3 + 2(\hat{Q}_2 - \hat{Q}_3)^2 \right]^{1 - L_1} \right| \leq E_{sf} + E_{sapp},$$

where, for
$$t \in \{2, 3\}$$

$$\begin{split} \hat{Q}_t &= \left(2\hat{Q}_{t,2} - \hat{Q}_{t,3}\right) \left[1 + \hat{Q}_{t,2} - \hat{Q}_{t,3} + 2(\hat{Q}_{t,2} - \hat{Q}_{t,3})^2\right]^{1-L_2} \\ E_{sf} &= (L_1 - 1)(L_2 - 1)\left(\beta_{2,2} + \beta_{2,3} + \beta_{3,2} + \beta_{3,3}\right) \\ E_{sapp} &= (L_1 - 1)\left[F_1\left(1 - \hat{Q}_2 + A_2 + C_2\right)^2 + (L_2 - 1)(F_2C_2 + F_3C_3)\right] \\ A_2 &= (L_2 - 1)\left(\beta_{2,2} + \beta_{2,3}\right) \\ C_t &= (L_2 - 1)F_t\left(1 - \hat{Q}_{t,2} + \beta_{t,2}\right)^2. \end{split}$$

ILLUSTRATION OF THE APPROXIMATION PROCESS



 $\begin{array}{c} \text{Find} \\ \text{Approximation} \end{array}$



ILLUSTRATION OF THE APPROXIMATION PROCESS

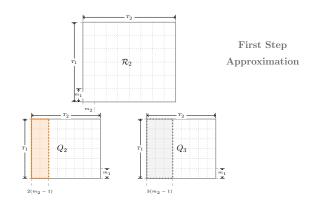
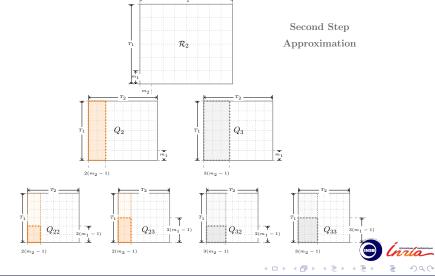




ILLUSTRATION OF THE APPROXIMATION PROCESS



METHODOLOGY SIMULATION METHODS: NORMAL DATA

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Simulation methods for Normal data





IMPORTANCE SAMPLING ALGORITHM

Test the null hypothesis of randomness against an alternative of clustering

 H_0 : The r.v.'s X_{s_1,s_2} are i.i.d. $\mathcal{N}(\mu,\sigma^2)$

 H_1 : There exists $\mathcal{R}(i_1, i_2) = [i_1 - 1, i_1 + m_1 - 1] \times [i_2 - 1, i_2 + m_2 - 1] \subset \mathcal{R}_2$ where the r.v.'s $X_{s_1,s_2} \sim \mathcal{N}(\mu_1,\sigma^2)$, $\mu_1 > \mu$ and $X_{s_1,s_2} \sim \mathcal{N}(\mu,\sigma^2)$ outside $\mathcal{R}(i_1,i_2)$

OBJECTIVE

Find a good estimate for $\mathbb{P}_{H_0}(S_m(T;\mathcal{S}) \geq \tau)$.

We are interested in evaluating the probability

$$\mathbb{P}_{\textit{H}_{0}}\left(\textit{S}_{m}(\textbf{T};\mathcal{S}) \geq \tau\right) = \mathbb{P}\left(\bigcup_{\textit{i}_{1}=1}^{\textit{T}_{1}-\textit{m}_{1}+1}\bigcup_{\textit{i}_{2}=1}^{\textit{T}_{2}-\textit{m}_{2}+1}\textit{E}_{\textit{i}_{1},\textit{i}_{2}}(\mathcal{S})\right)$$

where $E_{i_1,i_2}(S) = \{Y_{i_1,i_2}(S) > \tau\}$.



IMPORTANCE SAMPLING ALGORITHM

Algorithm 1 Importance Sampling Algorithm for Scan Statistics

Begin

Repeat for each k from 1 to ITER (iterations number)

- 1: Generate uniformly the couple $(i_1^{(k)}, i_2^{(k)})$ from the set $\{1, \ldots, T_1 m_1 + 1\} \times \{1, \ldots, T_2 m_1 + 1\}$ $m_2 + 1$.
- 2: Given the couple $(i_1^{(k)}, i_2^{(k)})$, generate a sample of the random field $\tilde{\mathbf{X}}^{(k)} = \{\tilde{X}_{s_1, s_2}^{(k)}\}$, with $s_j \in$ $\{1,\ldots,T_j\}$ and $j\in\{1,2\}$, from the conditional distribution of **X** given $\left\{Y_{i(k),j(k)}(\mathcal{S})\geq \tau\right\}$.
- 3: Take $c_k = C(\widetilde{\mathbf{X}}^{(k)})$ the number of all couples (i_1,i_2) for which $\widetilde{Y}_{i_1,i_2}(\mathcal{S}) \geq au$ and put $\widehat{\rho}_k(2) = \frac{1}{c_k}$

End Repeat Return

$$\widehat{\rho}(2) = \frac{1}{\mathit{ITER}} \sum_{k=1}^{\mathit{ITER}} \widehat{\rho}_k(2), \quad \mathit{Var}\left[\widehat{\rho}(2)\right] \approx \frac{1}{\mathit{ITER} - 1} \sum_{k=1}^{\mathit{ITER}} \left(\widehat{\rho}_k(2) - \frac{1}{\mathit{ITER}} \sum_{k=1}^{\mathit{ITER}} \widehat{\rho}_k(2)\right)^2$$

End



IMPORTANCE SAMPLING ALGORITHM: $\mathcal{N}(\mu, \sigma^2)$

Step 2 requires to sample:

- $Y_{i_i^{(k)},i_j^{(k)}}(\mathcal{S})$ from the tail distribution $\mathbb{P}\left(Y_{i_i^{(k)},i_j^{(k)}}(\mathcal{S}) \geq \tau\right)$ ([Devroye, 1986])
- ullet for the other indices, from the conditional distribution given $\left\{Y_{:(k)::(k)}(\mathcal{S})\geq au
 ight\}$

Lemma (Generalization of [Amarioarei, 2014, Lemma 3.4.4])

Let N be a positive integer, $\mathbf{X} = (X_1, X_2, \dots, X_N)$ be a vector of i.i.d. $\mathcal{N}(\mu, \sigma^2)$ and $\mathbf{a} = (a_1, \dots, a_N) \in \mathbb{R}^N$ a non zero constant vector ($a_i \neq 0$ for some particular j). Then conditionally given $\langle \mathbf{a}, \mathbf{X} \rangle = t$, the r.v.'s X_s with $s \neq j$ are jointly distributed as

$$\tilde{X}_{s} = \frac{a_{s}}{\|a\|} \left[\frac{t - \mu a_{j}}{\|a\|} - \frac{1}{\|a\| - |a_{j}|} \sum_{i \neq j} a_{i} \left(Z_{i} - \frac{\mu |a_{j}|}{\|a\|} \right) \right] + Z_{s}$$

where Z_s are i.i.d. $\mathcal{N}(\mu, \sigma^2)$ r.v.s.

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Numerical examples



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Table 1 : Numerical results for $\mathbb{P}(S \leqslant \tau)$: Triangle

Window's shape

Triangle $(m_1 = 14, m_2 = 18, Nt = 133, IS = 1e5, IA = 1e6)$



	× _{s1} , _{s2}	$\sim B(1, 0.01)$		$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total	
3	0.916397	0.918667	0.004333	59	0.863336	0.897101	0.004902	
4	0.997488	0.997483	0.000082	60	0.936684	0.948160	0.002010	
5	0.999947	0.999947	0.000002	61	0.971529	0.974938	0.000894	
6	0.999999	0.999999	0	62	0.987094	0.988279	0.000412	
7	0.999999	0.999999	0	63	0.994372	0.994664	0.000192	
8	1.000000	1.000000	0	64	0.997563	0.997643	0.000089	
9	1.000000	1.000000	0	65	0.998946	0.998971	0.000041	
10	1.000000	1.000000	0	66	0.999564	0.999567	0.000018	
11	1.000000	1.000000	0	67	0.999817	0.999820	0.000008	

	X _{s1} ,s	$_{2} \sim \mathcal{P}(0.25)$		$m{X_{s_1,s_2}} \sim \mathcal{N}(0,1)$			
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
59	0.739866	0.820277	0.011813	50	0.911170	0.934999	0.002737
60	0.871829	0.902957	0.004483	51	0.945219	0.957777	0.001655
61	0.939577	0.950977	0 001911	52	0.966494	0.972997	0.001026
62	0.971673	0.975649	0.000890	53	0.979944	0.982929	0.000644
63	0.987075	0.988206	0.000425	54	0.987840	0.989469	0.000406
64	0.994104	0.994492	0.000204	55	0.992768	0.993477	0.000257
65	0.997384	0.997452	0.000098	56	0.995667	0.996022	0.000162
66	0.998821	0.998855	0.000046	57	0.997412	0.997574	0.000102
67	0.999489	0.999490	0.000022	58	0.998509	0.998563	0.000063
						and the same of	



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Table 1 : Numerical results for $\mathbb{P}(S \leqslant \tau)$: Triangle

Window's shape

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	^s ₁ ,s ₂	$\sim B(1, 0.01)$		$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$				
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4	0.997488	0.997483	0.000082	60	0.936684	0.948160	0.002010	
5	0.999947	0.999947	0.000002	61	0.971529	0.974938	0.000894	
6	0.999999	0.999999	0	62	0.987094	0.988279	0.000412	
7	0.999999	0.999999	0	63	0.994372	0.994664	0.000192	
8	1.000000	1.000000	0	64	0.997563	0.997643	0.000089	
9	1.000000	1.000000	0	65	0.998946	0.998971	0.000041	
10	1.000000	1.000000	0	66	0.999564	0.999567	0.000018	
11	1.000000	1.000000	0	67	0.999817	0.999820	0.000008	

		$X_{s_1,s_2} \sim \mathcal{P}(0.25)$				$X_{s_1,s_2} \sim \mathcal{N}(0,1)$				
	τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total		
_	59	0.739866	0.820277	0.011813	50	0.911170	0.934999	0.002737		
	60	0.871829	0.902957	0.004483	51	0.945219	0.957777	0.001655		
	61	0.939577	0.950977	0.001911	52	0.966494	0.972997	0.001026		
	62	0.971673	0.975649	0.000890	53	0.979944	0.982929	0.000644		
	63	0.987075	0.988206	0.000425	54	0.987840	0.989469	0.000406		
	64	0.994104	0.994492	0.000204	55	0.992768	0.993477	0.000257		
	65	0.997384	0.997452	0.000098	56	0.995667	0.996022	0.000162		
	66	0.998821	0.998855	0.000046	57	0.997412	0.997574	0.000102		
	67	0.999489	0.999490	0.000022	58	0.998509	0.998563	0.000063		
							destr.			



SCANNING A REGION OF SIZE $T_1 \times T_2 = 250 \times 250$

Table 2: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Rectangle

Window's shape

Rectangle $(m_1 = 11, m_2 = 12, Nt = 132, IS = 1e5, IA = 1e6)$



	X _{s1} , _{s2}	$\sim B(1, 0.01)$		$X_{s_1,s_2} \sim B(5,0.05)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total	
3	0.947896	0.948226	0.002110	59	0.856993	0.857094	0.005485	
4	0.997983	0.997986	0.000058	60	0.919416	0.919589	0.002300	
5	0.999943	0.999943	0.000001	61	0.956420	0.956569	0.001024	
6	0.999999	0.999999	0	62	0.977142	0.977065	0.000471	
7	0.999999	0.999999	0	63	0.988228	0.988208	0.000220	
8	1.000000	1.000000	0	64	0.994130	0.994095	0.000103	
9	1.000000	1.000000	0	65	0.997121	0.997107	0.000048	
10	1.000000	1.000000	0	66	0.998610	0.998607	0.000022	
11	1.000000	1.000000	0	67	0.999342	0.999342	0.000010	

	X s1 , s	$\sim \mathcal{P}(0.25)$		$X_{s_1,s_2} \sim \mathcal{N}(0,1)$			
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
59	0.764372	0.764495	0.013285	50	0.865858	0.865484	0.004572
60	0.857872	0.859289	0.005320	51	0.904555	0.904805	0.002691
61	0 918972	0.918732	0.002307	52	0.933323	0.933206	0.001620
62	0.954682	0.954579	0.001059	53	0.953950	0.953807	0.000993
63	0.975271	0.975391	0.000502	54	0.968340	0.968483	0.000617
64	0.986996	0.986966	0.000242	55	0.978632	0.978687	0.000386
65	0.993240	0.993261	0.000117	56	0.985791	0.985752	0.000242
66	0.996557	0.996551	0.000056	57	0.990621	0.990585	0.000152
67	0.998290	0.998283	0.000027	58	0.993837	0.993815	0.000096



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Rectangle $(m_1 = 11, m_2 = 12, Nt = 132, IS = 1e5, IA = 1e6)$



			$\sim B(1, 0.01)$		$X_{s_1,s_2} \sim B(5,0.05)$			
	τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
	3	0.947896	0.948226	0.002110	59	0.856993	0.857094	0.005485
	4	0.997983	0.997986	0.000058	60	0.919416	0.919589	0.002300
	5	0.999943	0.999943	0.000001	61	0.956420	0.956569	0.001024
	6	0.999999	0.999999	0	62	0.977142	0.977065	0.000471
	7	0.999999	0.999999	0	63	0.988228	0.988208	0.000220
	8	1.000000	1.000000	0	64	0.994130	0.994095	0.000103
	9	1.000000	1.000000	0	65	0.997121	0.997107	0.000048
	10	1.000000	1.000000	0	66	0.998610	0.998607	0.000022
	11	1.000000	1.000000	0	67	0.999342	0.999342	0.000010
-								

	X _{s1} , _{s2}	$\sim \mathcal{P}(0.25)$		$X_{\mathbf{s_1},\mathbf{s_2}} \sim \mathcal{N}(0,1)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total	
59	0.764372	0.764495	0.013285	50	0.865858	0.865484	0.004572	
60	0.857872	0.859289	0.005320	51	0.904555	0.904805	0.002691	
61	0.918972	0.918732	0.002307	52	0.933323	0.933206	0.001620	
62	0.954682	0.954579	0 001059	53	0.953950	0.953807	0.000993	
63	0.975271	0.975391	0.000502	54	0.968340	0.968483	0.000617	
64	0.986996	0.986966	0.000242	55	0.978632	0.978687	0.000386	
65	0.993240	0.993261	0.000117	56	0.985791	0.985752	0.000242	
66	0.996557	0.996551	0.000056	57	0.990621	0.990585	0.000152	
67	0.998290	0.998283	0.000027	58	0.993837	0.993815	0.000096	



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Table 3: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Quadrilateral

Window's shape

Quadrilateral ($m_1 = 14, m_2 = 18, Nt = 131, IS = 1e5, IA = 1e6$)



	× _{s1} , _{s2}	$\sim B(1, 0.01)$		$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total	
3	0.926068	0.927398	0.003806	59	0.914546	0.927613	0.002942	
4	0.997622	0.997627	0.000075	60	0.959599	0.963873	0.001255	
5	0.999946	0.999946	0.000002	61	0.981235	0.982506	0.000571	
6	0.999999	0.999999	0	62	0 991423	0 991796	0.000266	
7	0.999999	0.999999	0	63	0.996113	0.996233	0.000124	
8	1.000000	1.000000	0	64	0.998283	0.998337	0.000057	
9	1.000000	1.000000	0	65	0.999266	0.999266	0.000026	
10	1.000000	1.000000	0	66	0.999684	0.999684	0.000012	
11	1.000000	1.000000	0	67	0.999868	0.999869	0.000005	

		X _{s1} ,s	$\sim \mathcal{P}(0.25)$		$X_{s_1,s_2} \sim \mathcal{N}(0,1)$			
	τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
•	59	0.835054	0.870351	0.006852	50	0.920004	0.935266	0.002571
	60	0.917972	0 931040	0.002768	51	0.950232	0 957711	0.001556
	61	0.960397	0.964711	0.001237	52	0.968755	0.972594	0.000964
	62	0 981228	0 982451	0.000585	53	0.980695	0.982566	0.000606
	63	0.991142	0.991510	0.000281	54	0.988110	0.989060	0.000383
	64	0.995855	0.995971	0.000136	55	0.992626	0.993110	0.000242
	65	0.998108	0.998124	0.000065	56	0.995569	0.995771	0.000153
	66	0.999135	0.999153	0.000031	57	0.997361	0.997394	0.000096
	67	0.999620	0.999622	0.000014	58	0.998379	0.998435	0.000060
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Scanning a region of size $T_1 \times T_2 = 250 \times 250$

Table 3: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Quadrilateral

Window's shape

Quadrilateral ($m_1 = 14, m_2 = 18, Nt = 131, IS = 1e5, IA = 1e6$)



	$\lambda_{s_1,s_2} \sim D(1,0.01)$				$\lambda_{s_1,s_2} \sim D(s,0.05)$				
_	τ	Sim	АррН	ETotal .	τ	Sim	АррН	E Total	
_	3	0.926068	0.927398	0.003806	59	0.914546	0.927613	0.002942	
	4	0.997622	0.997627	0.000075	60	0.959599	0.963873	0.001255	
	5	0.999946	0.999946	0.000002	61	0.981235	0.982506	0.000571	
	6	0.999999	0.999999	0	62	0.991423	0.991796	0.000266	
	7	0.999999	0.999999	0	63	0.996113	0.996233	0.000124	
	8	1.000000	1.000000	0	64	0.998283	0.998337	0.000057	
	9	1.000000	1.000000	0	65	0.999266	0.999266	0.000026	
	10	1.000000	1.000000	0	66	0.999684	0.999684	0.000012	
	11	1.000000	1.000000	0	67	0.999868	0.999869	0.000005	

		$X_{\mathbf{s_1},\mathbf{s_2}} \sim \mathcal{P}(0.25)$				$egin{aligned} X_{oldsymbol{s_1},oldsymbol{s_2}} &\sim \mathcal{N}(0,1) \end{aligned}$				
	τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total		
-	59	0.835054	0.870351	0.006852	50	0.920004	0.935266	0.002571		
	60	0.917972	0.931040	0.002768	51	0.950232	0.957711	0.001556		
	61	0.960397	0.964711	0.001237	52	0.968755	0.972594	0.000964		
	62	0.981228	0.982451	0.000585	53	0.980695	0.982566	0.000606		
	63	0.991142	0.991510	0.000281	54	0.988110	0.989060	0.000383		
	64	0.995855	0.995971	0.000136	55	0.992626	0 993110	0.000242		
	65	0.998108	0.998124	0.000065	56	0.995569	0.995771	0.000153		
	66	0.999135	0.999153	0.000031	57	0.997361	0.997394	0.000096		
	67	0.999620	0.999622	0.000014	58	0.998379	0.998435	0.000060		
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Table 4: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Circle

Window's shape

Circle $(m_1 = 13, m_2 = 13, Nt = 129, IS = 1e54, IA = 1e6)$



$X_{s_1,s_2} \sim B(1,0.01)$					$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$					
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total			
3	0.950311	0.950461	0.002195	59	0.920229	0.920388	0.002318			
4	0.998118	0.998114	0.000059	60	0.956814	0.957143	0.001016			
5	0.999947	0.999947	0.000001	61	0.977460	0.977614	0.000462			
6	0.999999	0.999999	0	62	0.988568	0.988567	0.000214			
7	0.999999	0.999999	0	63	0.994312	0.994309	0.000099			
8	1.000000	1.000000	0	64	0.997229	0.997228	0.000046			
9	1.000000	1.000000	0	65	0.998678	0.998679	0.000021			
10	1.000000	1.000000	0	66	0.999380	0.999381	0.000009			
11	1.000000	1.000000	0	67	0.999715	0.999715	0.000004			

		X_{s_1,s_2}	$_{2} \sim \mathcal{P}(0.25)$		$X_{s_1,s_2} \sim \mathcal{N}(0,1)$				
	τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total	
-	59	0.858454	0.859178	0.005391	50	0.888137	0.887891	0.003485	
	60	0.919182	0.919586	0.002310	51	0.921173	0.921549	0.002058	
	61	0.955229	0.955388	0.001047	52	0.945761	0.945644	0.001243	
	62	0.976023	0.975987	0.000491	53	0.962790	0.962825	0.000760	
	63	0.987414	0.987344	0.000234	54	0.974848	0.974878	0.000470	
	64	0.993477	0.993502	0.000112	55	0.983235	0.983263	0.000293	
	65	0.996706	0.996703	0.000054	56	0.988885	0.988907	0.000182	
	66	0.998372	0.998365	0.000025	57	0.992730	0.992734	0.000114	
	67	0.999207	0.999203	0.000012	58	0.995269	0.995287	0.000071	



Table 4: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Circle

Window's shape

Circle $(m_1 = 13, m_2 = 13, Nt = 129, IS = 1e54, IA = 1e6)$



$X_{s_1,s_2} \sim \mathcal{B}(1,0.01)$				$X_{s_1, s_2} \sim \mathcal{B}(5, 0.05)$					
τ	Sim	AppH	E Total	τ	Sim	АррН	ETotal		
3	0.950311	0.950461	0.002195	59	0.920229	0.920388	0.002318		
4	0.998118	0.998114	0.000059	60	0.956814	0.957143	0.001016		
5	0.999947	0.999947	0.000001	61	0.977460	0.977614	0.000462		
6	0.999999	0.999999	0	62	0.988568	0.988567	0.000214		
7	0.999999	0.999999	0	63	0.994312	0.994309	0.000099		
8	1.000000	1.000000	0	64	0.997229	0.997228	0.000046		
9	1.000000	1.000000	0	65	0.998678	0.998679	0.000021		
10	1 000000	1.000000	0	66	0.999380	0.999381	0.000009		
11	1.000000	1.000000	0	67	0.999715	0.999715	0.000004		

	X _{s1} ,	$_{5_{2}} \sim \mathcal{P}(0.25)$		$X_{s_1,s_2} \sim \mathcal{N}(0,1)$				
7		АррН	E Tot al	τ	Sim	АррН	ETotal	
5	9 0.858454	0.859178	0.005391	50	0.888137	0.887891	0.003485	
6	0 0.919182	0.919586	0.002310	51	0.921173	0.921549	0.002058	
6	1 0.955229	0.955388	0.001047	52	0.945761	0.945644	0.001243	
6	2 0.976023	0.975987	0.000491	53	0.962790	0.962825	0.000760	
6	3 0.987414	0.987344	0.000234	54	0.974848	0.974878	0.000470	
6	4 0.993477	0.993502	0.000112	55	0.983235	0.983263	0.000293	
6	5 0.996706	0.996703	0.000054	56	0.988885	0.988907	0.000182	
6	6 0.998372	0.998365	0.000025	57	0.992730	0.992734	0.000114	
6	7 0.999207	0.9992032	0.000012	58	0.995269	0.995287	0.000071	



Table 5: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Ellipse

Window's shape

Ellipse ($m_1 = 19, m_2 = 9, Nt = 135, IS = 1e5, IA = 1e6$)



$X_{s_1,s_2} \sim \mathcal{B}(1,0.01)$					$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total		
3	0.944001	0.944211	0.002297	59	0.764871	0.763482	0.009128		
4	0.997757	0.997758	0.000069	60	0.860903	0.860548	0.004127		
5	0.999935	0.999935	0.000002	61	0.921089	0.920882	0.001941		
6	0.999998	0.999998	0	62	0.956735	0.956866	0.000934		
7	1.000000	1.000000	0	63	0.977118	0.977094	0.000452		
8	1.000000	1.000000	0	64	0.988182	0.988152	0.000218		
9	1.000000	1.000000	0	65	0.994044	0.994012	0.000104		
10	1.000000	1.000000	0	66	0.997037	0.997037	0.000049		
11	1.000000	1.000000	0	67	0.998554	0.998558	0.000023		

	X _{s1} ,s	$_{2} \sim \mathcal{P}(0.25)$		$X_{s_1,s_2} \sim \mathcal{N}(0,1)$					
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total		
59	0.638962	0.639102	0.019442	50	0.843156	0.844113	0.004369		
60	0.768283	0.769666	0.008610	51	0.887477	0.887692	0.002755		
61	0 861614	0.860885	0.004012	52	0.920601	0.920385	0.001757		
62	0.919144	0.919301	0.001948	53	0.944398	0.944328	0.001127		
63	0.954941	0.954864	0.000965	54	0.961682	0.961667	0.000725		
64	0.975255	0.975369	0.000481	55	0.973797	0.973864	0.000468		
65	0.986869	0.986900	0.000240	56	0.982232	0.982330	0.000301		
66	0.993115	0.993127	0.000119	57	0.988183	0.988132	0.000193		
67	0.996485	0.996472	0.000058	58	0.992138	0.992134	0.000123		



Table 5: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Ellipse

Window's shape

Ellipse $(m_1 = 19, m_2 = 9, Nt = 135, IS = 1e5, IA = 1e6)$



$X_{s_1,s_2} \sim \mathcal{B}(1,0.01)$				$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total	
3	0.944001	0.944211	0.002297	59	0.764871	0.763482	0.009128	
4	0.997757	0.997758	0.000069	60	0.860903	0.860548	0.004127	
5	0.999935	0.999935	0.000002	61	0.921089	0.920882	0.001941	
6	0.999998	0.999998	0	62	0.956735	0.956866	0.000934	
7	1.000000	1.000000	0	63	0.977118	0.977094	0.000452	
8	1 000000	1.000000	0	64	0.988182	0.988152	0.000218	
9	1.000000	1.000000	0	65	0.994044	0.994012	0.000104	
10	1.000000	1.000000	0	66	0.997037	0.997037	0.000049	
11	1.000000	1.000000	0	67	0.998554	0.998558	0.000023	

$X_{s_1,s_2} \sim \mathcal{P}(0.25)$					$X_{s_1,s_2} \sim \mathcal{N}(0,1)$				
τ	Sim	АррН	ETotal	τ	Sim	АррН	ETotal		
59	0.638962	0.639102	0.019442	50	0.843156	0.844113	0.004369		
60	0.768283	0.769666	0.008610	51	0.887477	0.887692	0.002755		
61	0.861614	0.860885	0.004012	52	0.920601	0.920385	0.001757		
62	0 919144	0.919301	0.001948	53	0.944398	0.944328	0.001127		
63	0 954941	0.954864	0.000965	54	0 961682	0.961667	0.000725		
64	0.975255	0.975369	0.000481	55	0.973797	0.973864	0.000468		
65	0.986869	0.986900	0.000240	56	0.982232	0.982330	0.000301		
66	0.993115	0.993127	0.000119	57	0.988183	0.988132	0.000193		
67	0.996485	0.996472	0.000058	58	0.992138	0.992134	0.000123		
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Table 6: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Ellipse2

Window's shape

Ellipse2 ($m_1 = 9, m_2 = 19, Nt = 135, IS = 1e5, IA = 1e6$)



	^s ₁ ,s ₂	$\sim B(1, 0.01)$			^s ₁ ,s ₂	$\sim B(5, 0.05)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
3	0.944180	0.943779	0.003420	59	0.764090	0.763753	0.036701
4	0.997761	0.997767	0.000071	60	0.859511	0.860602	0.011219
5	0.999934	0.999935	0.000002	61	0.921173	0.920669	0.003816
6	0.999998	0.999998	0	62	0.956920	0.956693	0.001440
7	0.999999	0.999999	0	63	0.977198	0.977129	0.000586
8	1.000000	1.000000	0	64	0.988162	0.988177	0.000253
9	1.000000	1.000000	0	65	0.993979	0.994008	0.000113
10	1.000000	1.000000	0	66	0.997032	0.997036	0.000051
11	1.000000	1.000000	0	67	0.998558	0.998562	0.000023

	X _{s1} ,s	$_{2} \sim \mathcal{P}(0.25)$			X _{s1} ,s	$_{2} \sim \mathcal{N}(0,1)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
59	0.638389	0.640805	0.117483	50	0.843792	0.844342	0.013346
60	0.770899	0.769795	0.034030	51	0.888070	0.887823	0.006857
61	0.860847	0.861092	0.010890	52	0.920460	0.920691	0.003626
62	0.919522	0.919537	0.003909	53	0.944514	0.944368	0.001974
63	0.954873	0.954742	0.001516	54	0 961591	0.961748	0.001109
64	0.975326	0.975379	0.000635	55	0.973861	0.973830	0.000640
65	0.986856	0.986843	0.000282	56	0.982294	0.982269	0.000377
66	0.993154	0.993119	0.000130	57	0.988180	0.988143	0.000226
67	0.996482	0.996478	0.000061	58	0.992154	0.992123	0.000138
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Scanning a region of size $T_1 \times T_2 = 250 \times 250$

Table 6: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Ellipse2

Window's shape

Ellipse2 ($m_1 = 9, m_2 = 19, Nt = 135, IS = 1e5, IA = 1e6$)



	^s ₁ ,s ₂	$\sim B(1,0.0)$			^ s ₁ , s ₂	$\sim B(5, 0.05)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
3	0.944180	0.943779	0.003420	59	0.764090	0.763753	0.036701
4	0.997761	0.997767	0.000071	60	0.859511	0.860602	0.011219
5	0.999934	0.999935	0.000002	61	0.921173	0.920669	0.003816
6	0.999998	0.999998	0	62	0.956920	0.956693	0 001440
7	0.999999	0.999999	0	63	0.977198	0.977129	0.000586
8	1.000000	1.000000	0	64	0.988162	0 988177	0.000253
9	1.000000	1.000000	0	65	0.993979	0.994008	0.000113
10	1.000000	1.000000	0	66	0.997032	0.997036	0.000051
11	1.000000	1.000000	0	67	0.998558	0.998562	0.000023

	X_{s_1, s_2}	$\sim \mathcal{P}(0.25)$			X ₅₁ ,5	$_{2} \sim \mathcal{N}(0, 1)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	ETotal
59	0.638389	0.640805	0.117483	50	0.843792	0.844342	0.013346
60	0.770899	0.769795	0.034030	51	0.888070	0.887823	0.006857
61	0.860847	0.861092	0.010890	52	0.920460	0.920691	0.003626
62	0 919522	0.919537	0.003909	53	0.944514	0.944368	0.001974
63	0.954873	0.954742	0.001516	54	0 961591	0 961748	0.001109
64	0.975326	0.975379	0.000635	55	0.973861	0.973830	0.000640
65	0.986856	0.986843	0.000282	56	0.982294	0.982269	0.000377
66	0.993154	0.993119	0.000130	57	0.988180	0.988143	0.000226
67	0.996482	0.996478	0.000061	58	0.992154	0.992123	0.000138
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Table 7: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Annulus

Window's shape

Annulus ($m_1 = 17, m_2 = 17, Nt = 124, IS = 1e5, IA = 1e6$)



	X _{s1} , _{s2}	$\sim B(1, 0.01)$			X s ₁ , s ₂	$\sim \mathcal{B}(5, 0.05)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
3	0.881798	0.882489	0.004812	59	0.951170	0.951245	0.000699
4	0.995434	0.995465	0.000069	60	0.975772	0.975727	0.000255
5	0.999883	0.999883	0.000001	61	0 988275	0.988270	0.000099
6	0.999998	0.999998	0	62	0.994460	0.994466	0.000041
7	0.999999	0.999999	0	63	0.997439	0.997440	0.000017
8	1.000000	1.000000	0	64	0.998839	0.998840	0.000007
9	1.000000	1.000000	0	65	0.999484	0.999484	0.000003
10	1.000000	1.000000	0	66	0.999775	0.999775	0.000001
11	1.000000	1.000000	0	67	0.999903	0.999903	0.000000

	X _{s1} ,s	$\sim \mathcal{P}(0.25)$			X _{s1} ,s	$_{2} \sim \mathcal{N}(0,1)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
59	0.903956	0.903852	0.002128	50	0.860708	0.860416	0.004097
60	0.949083	0.949059	0.000735	51	0 904651	0.904644	0.001977
61	0.973814	0.973844	0.000277	52	0.936077	0.935938	0.000987
62	0.986909	0.986876	0.000111	53	0.957564	0.957650	0.000508
63	0.993576	0.993570	0.000047	54	0.972378	0.972311	0.000270
64	0.996910	0.996907	0.000020	55	0.982139	0.982136	0.000148
65	0.998539	0.998539	0.000009	56	0.988564	0.988587	0.000082
66	0.999320	0.999320	0.000004	57	0.992769	0.992769	0.000047
67	0.999689	0.999689	0.000002	58	0.995471	0.995466	0.000027
						- 47000 p.	



Table 7: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Annulus

Window's shape

Annulus ($m_1 = 17, m_2 = 17, Nt = 124, IS = 1e5, IA = 1e6$)



	^s ₁ ,s ₂	$\sim B(1, 0.01)$			^s ₁ ,s ₂	$\sim B(5, 0.05)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	E Total
3	0.881798	0.882489	0.004812	59	0.951170	0 951245	0.000699
4	0.995434	0.995465	0.000069	60	0.975772	0.975727	0.000255
5	0.999883	0.999883	0.000001	61	0.988275	0.988270	0.000099
6	0.999998	0.999998	0	62	0.994460	0.994466	0.000041
7	0.999999	0.999999	0	63	0.997439	0.997440	0.000017
8	1.000000	1.000000	0	64	0.998839	0.998840	0.000007
9	1.000000	1.000000	0	65	0.999484	0.999484	0.000003
10	1.000000	1.000000	0	66	0.999775	0.999775	0.000001
11	1.000000	1.000000	0	67	0.999903	0.999903	0.000000

	X_{s_1, s_2}	$\sim \mathcal{P}(0.25)$			X ₅₁ ,5	$_{2} \sim \mathcal{N}(0,1)$	
τ	Sim	АррН	ETotal	τ	Sim	АррН	ETotal
59	0.903956	0.903852	0.002128	50	0.860708	0.860416	0.004097
60	0.949083	0.949059	0.000735	51	0.904651	0.904644	0.001977
61	0.973814	0.973844	0.000277	52	0.936077	0.935938	0.000987
62	0.986909	0.986876	0.000111	53	0.957564	0.957650	0.000508
63	0.993576	0.993570	0.000047	54	0.972378	0.972311	0.000270
64	0.996910	0.996907	0.000020	55	0.982139	0 982136	0.000148
65	0.998539	0.998539	0.000009	56	0.988564	0.988587	0.000082
66	0.999320	0.999320	0.000004	57	0.992769	0.992769	0.000047
67	0.999689	0.999689	0.000002	58	0.995471	0.995466	0.000027



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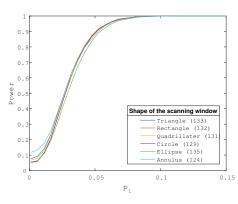


Power of the scan statistic test

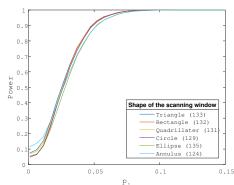


Power evaluation for $\mathcal{B}(1,0.001)$ model

Triangular simulated cluster



Rectangular simulated cluster

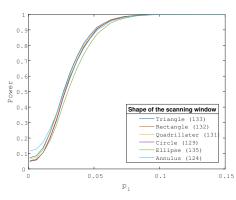




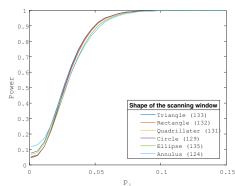


Power evaluation for $\mathcal{B}(1,0.001)$ model

Quadrilateral simulated cluster



Circular simulated cluster

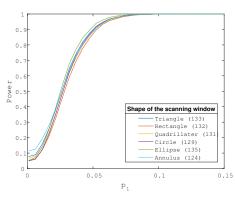




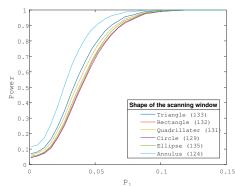


Power evaluation for $\mathcal{B}(1,0.001)$ model

Ellipsoidal simulated cluster



Annular simulated cluster

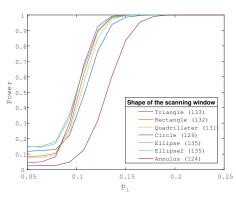




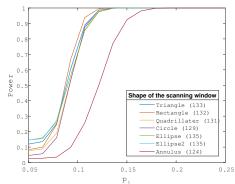


Power evaluation for $\mathcal{B}(5,0.05)$ model

Triangular simulated cluster



Rectangular simulated cluster

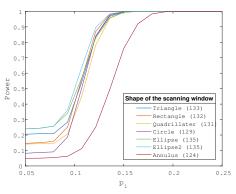




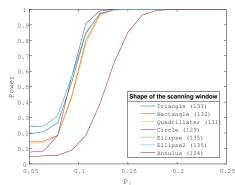


Power evaluation for $\mathcal{B}(5,0.05)$ model

Quadrilateral simulated cluster



Circular simulated cluster

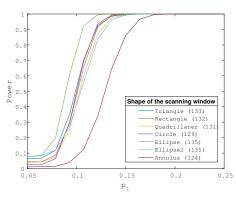




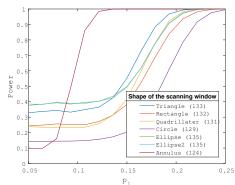


Power evaluation for $\mathcal{B}(5,0.05)$ model

Ellipsoidal simulated cluster



Annular simulated cluster







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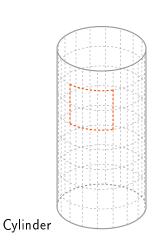
Scanning the surface of a cylinder





IWAP 2016

SCANNING THE SURFACE OF A CYLINDER



Unfolded cylinder of size $T_1 \times T_2$

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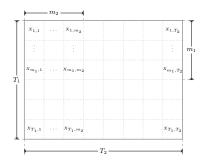




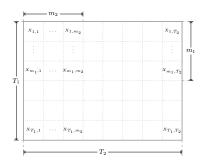
Transformation into a one dimensional problem





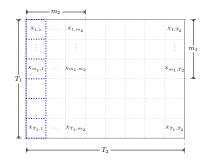






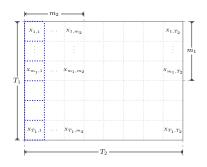






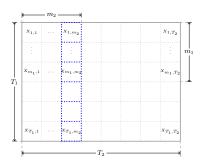




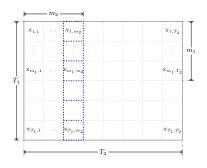




IWAP 2016

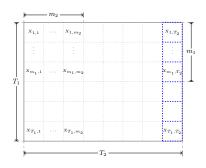






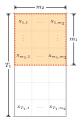
			X_{T_1,m_2}			
A1.1						





$X_{1,1}$	X_{m_1}	X_{T_1}	,1	$X_{1,i}$	n ₂	x_{m_1}	m ₂	X_{T_1}	,m ₂	$X_{1,T}$	 X_{m_1,T_2}	$X_{T_{1},T_{2}}$





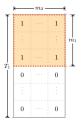
- The size of the scanning window is $\tilde{m}_1 = T_1 m_2 (T_1 m_1)$
- The score function is defined by

$$\mathcal{S}\left(\mathfrak{X}_{i_{1}}\right)=A\circ\mathfrak{X}_{i_{1}}$$

where A is the corresponding $\{0,1\}$ vector



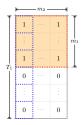
IWAP 2016



- The size of the scanning window is $\tilde{m}_1 = T_1 m_2 (T_1 m_1)$
- The score function is defined by

$$\mathcal{S}\left(\mathfrak{X}_{i_{1}}\right)=A\circ\mathfrak{X}_{i_{1}}$$



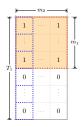


- The size of the scanning window is $\tilde{m}_1 = T_1 m_2 - (T_1 - m_1)$
- The score function is defined by

$$\mathcal{S}\left(\mathfrak{X}_{i_{1}}\right)=A\circ\mathfrak{X}_{i_{1}}$$





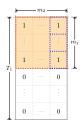


- The size of the scanning window is $\tilde{m}_1 = T_1 m_2 (T_1 m_1)$
- The score function is defined by

$$\mathcal{S}\left(\mathfrak{X}_{i_{1}}\right)=A\circ\mathfrak{X}_{i_{1}}$$





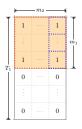


- The size of the scanning window is $\tilde{m}_1 = T_1 m_2 (T_1 m_1)$
- The score function is defined by

$$S\left(\mathfrak{X}_{i_1}\right) = A \circ \mathfrak{X}_{i_1}$$







- The size of the scanning window is $\tilde{m}_1 = T_1 m_2 (T_1 m_1)$
- The score function is defined by

$$\mathcal{S}\left(\mathfrak{X}_{i_{1}}\right)=A\circ\mathfrak{X}_{i_{1}}$$





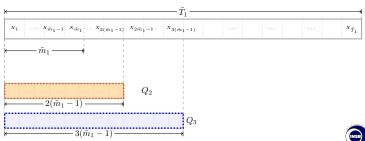
APPROXIMATION AND ERROR BOUNDS

Theorem [Amărioarei, 2014]

Let $t_1 \in \{2,3\}$ and $Q_{t_1} = Q_{t_1}(\tau) = \mathbb{P}(S_{\tilde{m}_1}(t_1(\tilde{m}_1 - 1); S) \leq \tau)$ and $L_1 = \left| \frac{T_1}{\tilde{m}_1 - 1} \right|$ If \hat{Q}_{t_1} is an estimate of Q_{t_1} with $\left|\hat{Q}_{t_1}-Q_{t_1}\right|\leq eta_{t_1}$ and au is such that $1-\hat{Q}_2(au)\leq 0.1$ then

$$\left| \mathbb{P}\left(S_{\tilde{m}_{1}}(\tilde{T}_{1}, \mathcal{S}) \leq \tau \right) - \left(2\hat{Q}_{2} - \hat{Q}_{3} \right) \left[1 + \hat{Q}_{2} - \hat{Q}_{3} + 2(\hat{Q}_{2} - \hat{Q}_{3})^{2} \right]^{1 - L_{1}} \right| \leq E_{total}(1),$$

$$E_{total}(1) = (L_{1} - 1) \left[\beta_{2} + \beta_{3} + F\left(\hat{Q}_{2}, L_{1} - 1 \right) \left(1 - \hat{Q}_{2} + \beta_{2} \right)^{2} \right].$$





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Numerical examples



Table 8: Numerical results for $\mathbb{P}(S\leqslant au)$: Cylinder

	$X_{s_1,s_2} \sim \mathcal{B}(1,0.1)$			$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$			
$\overline{\tau}$	Sim	AppH	E Tot al	au	Sim	AppH	E Tot a
33	0.871559	0.870200	0.003674	68	0.955593	0.955671	0.000938
34	0.946216	0.946527	0.001177	69	0.976348	0.976285	0.000461
35	0.979458	0.979381	0.000393	70	0.987406	0.987574	0.000227
36	0.992379	0.992465	0.000131	71	0.993526	0.993593	0.000111
37	0.997384	0.997367	0.000043	72	0.996772	0.996751	0.000054
38	0.999116	0.999111	0.000014	73	0.998401	0.998391	0.000026
39	0.999713	0.999714	0.000004	74	0.999214	0.999212	0.000012
40	0.999911	0.999911	0.000001	75	0.999623	0.999626	0.000006



Table 8: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Cylinder

	$X_{\mathbf{s_1},\mathbf{s_2}} \sim \mathcal{B}(1,0.1)$			$X_{s_1,s_2} \sim \mathcal{B}(5,0.05)$			
$\overline{\tau}$	Sim	AppH	ETot al	au	Sim	AppH	E Tot al
33	0.871559	0.870200	0.003674	68	0.955593	0.955671	0.000938
34	0.946216	0.946527	0.001177	69	0.976348	0.976285	0.000461
35	0.979458	0.979381	0.000393	70	0.987406	0.987574	0.000227
36	0.992379	0.992465	0.000131	71	0.993526	0.993593	0.000111
37	0.997384	0.997367	0.000043	72	0.996772	0.996751	0.000054
38	0.999116	0.999111	0.000014	73	0.998401	0.998391	0.000026
39	0.999713	0.999714	0.000004	74	0.999214	0.999212	0.000012
40	0.999911	0.999911	0.000001	75	0.999623	0.999626	0.000006



Table 9: Numerical results for $\mathbb{P}(S\leqslant au)$: Cylinder

$X_{s_1,s_2} \sim \mathcal{P}(0.25)$						
$\overline{\tau}$	Sim	AppH	ETot al			
68	0.915283	0.915691	0.002055			
69	0.951447	0.951723	0.001023			
70	0.973445	0.973488	0.000515			
71	0.985486	0.985509	0.000263			
72	0.992349	0.992285	0.000133			
73	0.995950	0.995979	0.000066			
74	0.997916	0.997920	0.000033			
75	0.998951	0.998945	0.000016			



Table 9: Numerical results for $\mathbb{P}(S \leqslant \tau)$: Cylinder

$X_{s_1,s_2} \sim \mathcal{P}(0.25)$						
au	Sim	AppH	ETot al			
68	0.915283	0.915691	0.002055			
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71	0.985486	0.985509	0.000263			
72	0.992349	0.992285	0.000133			
73	0.995950	0.995979	0.000066			
74	0.997916	0.997920	0.000033			
75	0.998951	0.998945	0.000016			



thank you!





Alm, S. E. (1983).

On the distribution of scan statistic of poisson process.

In Gut, A. and Helst, L., editors, *Probability and Mathematical Statistics*, pages 1–10. Upsalla University Press.



Alm, S. E. (1997).

On the distributions of scan statistics of a two-dimensional poisson process. *Advances in Applied Probability*, pages 1–18.



Alm, S. E. (1998).

Approximation and simulation of the distributions of scan statistics for poisson processes in higher dimensions.

Extremes, 1(1):111-126.



Amărioarei, A. (2014).

Approximations for the multidimensional discrete scan statistics.

PhD thesis, University of Lille 1.



Anderson, N. H. and Titterington, D. M. (1997).

Some methods for investigating spatial clustering, with epidemiological applications.

Journal of the Royal Statistical Society: Series A (Statistics in Society)



Assunção, R., Costa, M., Tavares, A., and Ferreira, S. (2006).

Fast detection of arbitrarily shaped disease clusters.

Statistics in Medicine, 25(5):723-742.



Bresenham, J. (1965).

Algorithm for computer control of a digital printer.

IBM Systems Journal, 4(1).



Bresenham, J. (1977).

A linear algorithm for incremental digital display of circular arcs.

Communications of the ACM, 20(2):100-106.



Devroye, L. (1986).

Non uniform random variate generation.

Springer-Verlag, New York.



Foley, J. (1995).

Computer Graphics, Principles and Practice in C.

Addison-Wesley Professional.



Glaz, J., Naus, J., and Wallenstein, S. (2001). *Scan Statistics*.



Springer.



Glaz, J., Pozdnyakov, V., and Wallenstein, S. (2009).

Scan Statistics: Methods and Applications.

Birkhaüser Boston.



Loader, C. R. (1991).

Large-deviation approximations to the distribution of scan statistics.

Advances in Applied Probability, pages 751-771.



Tango, T. and Takahashi, K. (2005).

A flexibly shaped spatial scan statistic for detecting clusters.

Int J Health Geogr, 4:11.

Tango, Toshiro Takahashi, Kunihiko England Int J Health Geogr. 2005 May 18;4:11.

