

Approximations for two-dimensional discrete scan statistics in some dependent models

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Outline

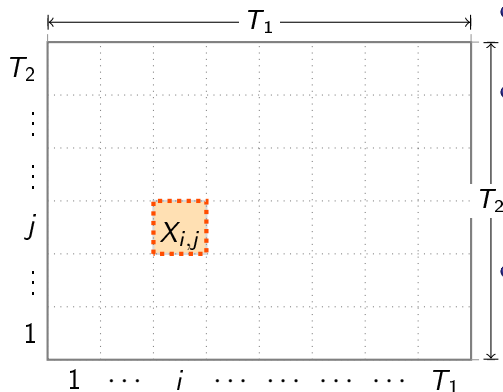
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 - Framework and Model
 - Previous Work
- 2 Description of the method
 - Main Idea and Tools
 - The Approximation
- 3 Error Bound
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Introducing the General Model

Let T_1, T_2 be positive integers



- Rectangular region
 $\mathcal{R} = [0, T_1] \times [0, T_2]$
- $(X_{ij})_{\substack{1 \leq i \leq T_1 \\ 1 \leq j \leq T_2}}$ integer r.v.'s
 - Bernoulli($\mathcal{B}(1, p)$)
 - Binomial($\mathcal{B}(n, p)$)
 - Poisson($\mathcal{P}(\lambda)$)
- X_{ij} number of observed events in the elementary subregion
 $r_{ij} = [i - 1, i] \times [j - 1, j]$

Introducing the Block-Factor Model

Consider for $1 \leq i \leq T_1, 1 \leq j \leq T_2$ the following block-factor model:

$$X_{i,j} = f(Y_{i,j}, Y_{i,j-1}, Y_{i,j+1}, Y_{i-1,j-1}, Y_{i-1,j}, Y_{i-1,j+1}, Y_{i+1,j-1}, Y_{i+1,j}, Y_{i+1,j+1}),$$

with $f: \mathbb{R}^9 \rightarrow \mathbb{R}_+$ and i.i.d. sequence

$$\{Y_{i,j} \mid 0 \leq i \leq T_1 + 1, 0 \leq j \leq T_2 + 1\}$$

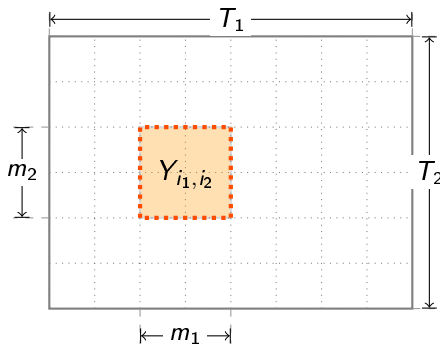
Y_{0,T_2+1}	Y_{1,T_2+1}	Y_{T_1+1,T_2+1}
.									.
.									.
.									.
.				$Y_{i-1,j+1}$	$Y_{i,j+1}$	$Y_{i+1,j+1}$.
.				$Y_{i-1,j}$	$Y_{i,j}$	$Y_{i+1,j}$.
.				$Y_{i-1,j-1}$	$Y_{i,j-1}$	$Y_{i+1,j-1}$.
.									.
.									.
$Y_{0,1}$									$Y_{T_1+1,1}$
$Y_{0,0}$	$Y_{1,0}$	$Y_{T_1+1,0}$

f

X_{1,T_2}	X_{2,T_2}	X_{T_1,T_2}
.							.
.							.
.							.
							$X_{i,j}$
.							.
.							.
.							.
$X_{1,2}$							$X_{T_1,2}$
$X_{1,1}$	$X_{2,1}$	$X_{T_1,1}$

Defining the Scan Statistic

Let m_1, m_2 be positive integers



- Define for $1 \leq i_j \leq T_j - m_j + 1$,

$$Y_{i_1 i_2} = \sum_{i=i_1}^{i_1+m_1-1} \sum_{j=i_2}^{i_2+m_2-1} X_{ij}$$

- The two dimensional scan statistic,

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

- Used for testing the null hypotheses of randomness against the alternative hypothesis of clustering

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Problem and related results

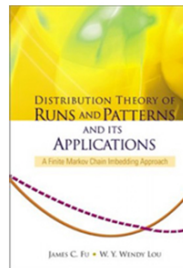
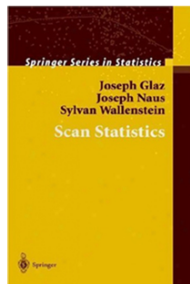
Problem

Approximate the distribution of two dimensional discrete scan statistic for the block-factor model

$$\mathbb{P}(S_{m_1, m_2}(T_1, T_2) \leq n).$$

- Dependent model: **no results !**
- Independent model:
 - No exact formulas
 - For Bernoulli case:
 - product type approximations (Boutsikas and Koutras 2000)
 - Poisson approximations (Chen and Glaz 1996)
 - bounds (Boutsikas and Koutras 2003)
 - For binomial and Poisson cases: (Glaz 2009)
 - Product type approximation
 - Lower bound

Literature



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Key Idea

Haiman(2000) proposed a different approach

Main Observation

The scan statistic r.v. can be viewed as a maximum of a sequence of 1-dependent stationary r.v..

- The idea:
 - discrete and continuous one dimensional scan statistic: Haiman (2000,2007)
 - discrete and continuous two dimensional scan statistic: Haiman and Preda (2002,2006)
 - discrete three dimensional scan statistic: Amarioarei (2013)

Writing the Scan as an Extreme of 1-Dependent R.V.'s

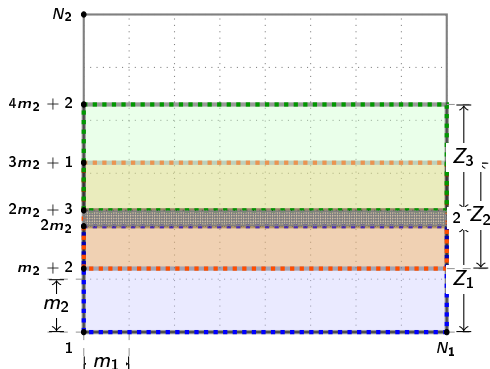
Let $T_j = (L_j + 1)(m_j + 1) - 2$,
 $j \in \{1, 2\}$ positive integers

- Define for $l \in \{1, 2, \dots, L_2\}$

$$Z_l = \max_{\substack{1 \leq i_1 \leq L_1(m_1+1) \\ (l-1)(m_2+1)+1 \leq i_2 \leq l(m_2+1)}} Y_{i_1 i_2}$$

- $(Z_l)_l$ is 1-dependent and stationary
- Observe

$$S_{m_1, m_2}(T_1, T_2) = \max_{1 \leq l \leq L_2} Z_l$$



Main Tool

Let $(Z_j)_{j \geq 1}$ be a strictly stationary 1-dependent sequence of r.v.'s and let $q_m = q_m(x) = \mathbb{P}(\max(Z_1, \dots, Z_m) \leq x)$, with $x < \sup\{u | \mathbb{P}(Z_1 \leq u) < 1\}$.

Main Theorem (Haiman 1999, Amarioarei 2012)

For x such that $\mathbb{P}(Z_1 > x) = 1 - q_1 \leq \alpha < 0.1$ and $m > 3$ we have

$$\left| q_m - \frac{6(q_1 - q_2)^2 + 4q_3 - 3q_4}{(1 + q_1 - q_2 + q_3 - q_4 + 2q_1^2 + 3q_2^2 - 5q_1q_2)^m} \right| \leq \Delta_1(1 - q_1)^3,$$

$$\left| q_m - \frac{2q_1 - q_2}{[1 + q_1 - q_2 + 2(q_1 - q_2)^2]^m} \right| \leq \Delta_2(1 - q_1)^2,$$

- $\Delta_1 = \Delta_1(\alpha, q_1, m) = \Gamma(\alpha) + mK(\alpha)$
- $\Delta_2 = mE(\alpha, q_1, m) = m \left[1 + \frac{3}{m} + K(\alpha)(1 - q_1) + \frac{\Gamma(\alpha)(1 - q_1)}{m} \right].$

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First Step Approximation

Using Main Theorem we obtain

- Define

$$Q_2 = \mathbb{P}(Z_1 \leq k)$$

$$Q_3 = \mathbb{P}(Z_1 \leq k, Z_2 \leq k)$$

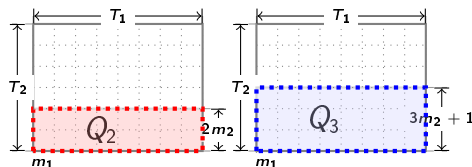
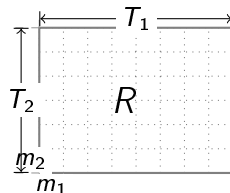
- If $1 - Q_2 \leq \alpha_1 < 0.1$ the (first) approximation

$$\mathbb{P}(S \leq k) \approx \frac{2Q_2 - Q_3}{[1 + Q_2 - Q_3 + 2(Q_2 - Q_3)^2]^{L_2}}$$

where $S = S_{m_1, m_2}(T_1, T_2)$

- Approximation error

$$L_2 E(\alpha_1, L_2)(1 - Q_2)^2$$



Second Step Approximation

Q_2 :

- For $s \in \{1, 2, \dots, L_1\}$

$$Z_s^{(2)} = \max_{\substack{(s-1)(m_1+1)+1 \leq i_1 \leq s(m_1+1) \\ 1 \leq i_2 \leq m_2+1}} Y_{i_1 i_2}$$

- $Q_2 = \mathbb{P} \left(\max_{1 \leq s \leq L_1} Z_s^{(2)} \leq k \right)$

- Define $Q_{22} = \mathbb{P}(Z_1^{(2)} \leq k)$
 $Q_{32} = \mathbb{P}(Z_1^{(2)} \leq k, Z_2^{(2)} \leq k)$

- Approximation ($1 - Q_{22} \leq \alpha_2$)

$$Q_2 \approx \frac{2Q_{22} - Q_{32}}{[1 + Q_{22} - Q_{32} + 2(Q_{22} - Q_{32})^2]^{L_1}}$$

- Error

$$L_1 E(\alpha_2, L_1)(1 - Q_{22})^2$$

Q_3 :

- For $s \in \{1, 2, \dots, L_1\}$

$$Z_s^{(3)} = \max_{\substack{(s-1)(m_1+1)+1 \leq i_1 \leq s(m_1+1) \\ 1 \leq i_2 \leq 2(m_2+1)}} Y_{i_1 i_2}$$

- $Q_3 = \mathbb{P} \left(\max_{1 \leq l \leq L_1} Z_s^{(3)} \leq k \right)$

- Define $Q_{23} = \mathbb{P}(Z_1^{(3)} \leq k)$
 $Q_{33} = \mathbb{P}(Z_1^{(3)} \leq k, Z_2^{(3)} \leq k)$

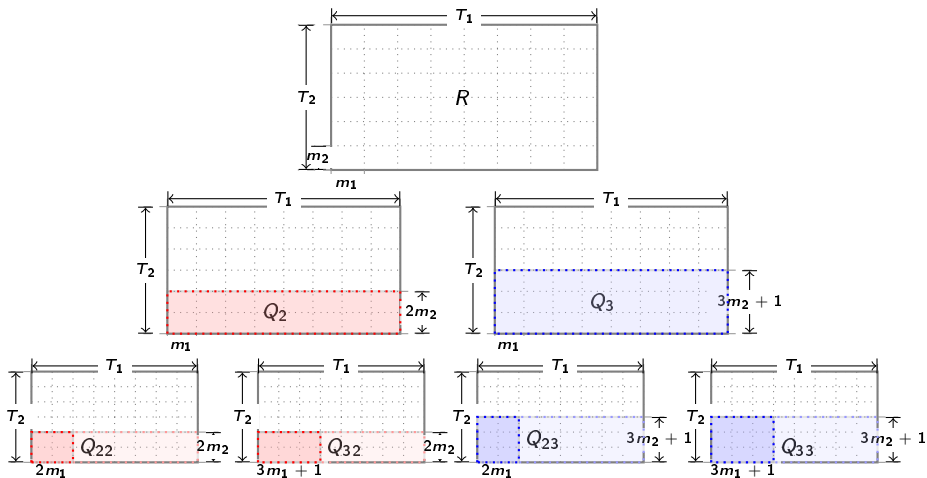
- Approximation ($1 - Q_{23} \leq \alpha_2$)

$$Q_3 \approx \frac{2Q_{23} - Q_{33}}{[1 + Q_{23} - Q_{33} + 2(Q_{23} - Q_{33})^2]^{L_1}}$$

- Error

$$L_1 E(\alpha_2, L_1)(1 - Q_{23})^2$$

Illustration of the Approximation Process



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Theoretical Approximation Error

Define for $s \in \{2, 3\}$

$$H(x, y, m) = \frac{2x - y}{[1 + x - y + 2(x - y)^2]^m}, \quad \alpha_1 = 1 - Q_3, \quad \alpha_2 = 1 - Q_{23},$$

$$E_1 = E(\alpha_2, L_1), \quad E_2 = E(\alpha_1, L_2), \quad R_s = H(Q_{2s}, Q_{3s}, L_1),$$

The approximation error

$$E_{app} = L_2 F_2 B_2^2 + L_1 L_2 F_1 [(1 - Q_{22})^2 + (1 - Q_{23})^2]$$

where B_2 is given by

$$B_2 = 1 - R_2 + L_1 F_1 (1 - Q_{22})^2$$

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Simulation Error for Approximation Formula

If $ITER$ is the number of simulations, we can say, at 95% confidence level,

$$\left| Q_{rt} - \hat{Q}_{rt} \right| \leq 1.96 \sqrt{\frac{\hat{Q}_{rt}(1-\hat{Q}_{rt})}{ITER}} = \beta_{rt}, \quad r, t \in \{2, 3\}$$

where \hat{Q}_{rt} is the simulated value.

Define for $r \in \{2, 3\}$,

$$\hat{Q}_r = H\left(\hat{Q}_{2r}, \hat{Q}_{3r}, L_1\right)$$

The simulation error corresponding to the approximation formula

$$E_{sf} = L_1 L_2 (\beta_{22} + \beta_{23} + \beta_{32} + \beta_{33})$$

Simulation Error for Approximation Error

Introducing

$$C_{2r} = 1 - \hat{Q}_{2r} + \beta_{2r}, \quad r \in \{2, 3\},$$

$$C_2 = 1 - \hat{Q}_2 + L_1(\beta_{22} + \beta_{32}) + L_1 F_1 C_{22}^2,$$

The simulation error corresponding to the approximation

$$E_{sapp} = L_2 F_2 C_2^2 + L_1 L_2 F_1 [C_{22}^2 + C_{23}^2]$$

The total error

$$E_{total} = E_{app} + E_{sf} + E_{sapp}$$

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Example Model

Consider for each $1 \leq i \leq T_1$ and $1 \leq j \leq T_2$:

$$X_{ij} = \begin{cases} 1, & \text{if } Y_{ij} = 1 \text{ and } \sum_{k \in \{-1,0,1\}} Y_{i+k,j+k} \geq 2, \\ 0, & \text{otherwise.} \end{cases}$$

- X_{ij} 's are dependent Bernoulli r.v.'s with parameter

$$p' = p [1 - (1 - p)^8]$$

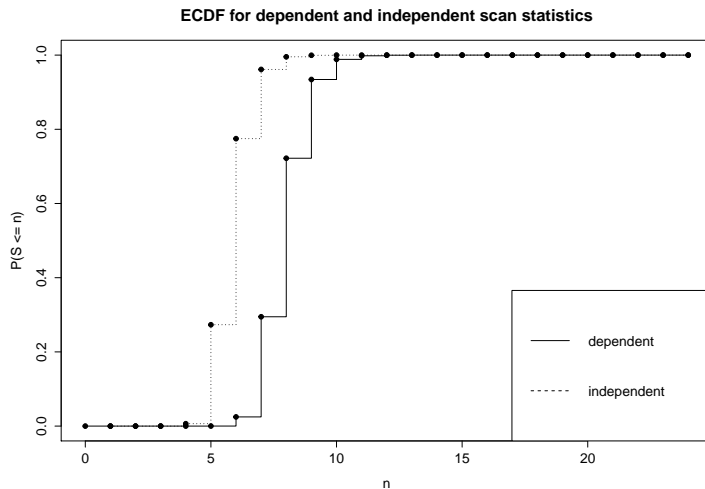
- $X_{ij} = 1$ each time when $Y_{ij} = 1$ and there is at least one success in its neighborhood (horizon one)







Numerical Results







Table 1 : $\mathbb{P}(S_{m_1, m_2}(T_1, T_2) \leq n)$: $m_1 = 4, m_2 = 6, T_1 = 53, T_2 = 75, ITER = 10^9$

n	Sim Dep	$Approx$ Dep	E_{app}	E_{sim}	E_{total}	Sim $Indep$	$Approx$ $Indep$
$p = 0.01, p' = 0.00077$							
2	0.91937	0.91959	0.00351	0.00167	0.00518	0.99956	0.99921
3	0.98750	0.98748	0.00004	0.00046	0.00051	1	0.99999
4	0.99930	0.99915	0.00000	0.00010	0.00010	1	1
5	0.99993	0.99993	0.00000	0.00002	0.00002	1	1
$p = 0.1, p' = 0.05695$							
9	0.93423	0.93247	0.00120	0.00111	0.00231	0.99957	0.99941
10	0.98847	0.98780	0.00003	0.00042	0.00045	0.99999	0.99995
11	0.99815	0.99812	0.00000	0.00015	0.00015	1	1
12	0.99971	0.99984	0.00000	0.00004	0.00004	1	1
13	0.99996	0.99999	0.00000	0.00001	0.00001	1	1

Graphical Illustration



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