

## Computer generation of initial spatial distribution for cell automata

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Received 8 May 2011; Accepted 13 June 2011; Published online 1 December 2011

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### Abstract

The algorithm to generate spatial distribution patterns was developed and presented in this study. Three typical spatial distribution patterns, i.e., Poisson distribution, binomial distribution, and negative binomial distribution were included in the algorithm. The Java program was also provided. The algorithm can be used to generate initial distribution in cell automata modeling.

**Keywords** spatial distribution patterns; computer generation; cell automata.

### 1 Introduction

Spatial distribution pattern refers mainly to the two-dimensional distribution of individuals in a region. There are many types of spatial distribution patterns (Krebs, 1999; Zhang, 2007). They can be classified into three categories, namely random distribution, uniform distribution, and aggregated distribution, which can be represented by Poisson distribution, binomial distribution and negative binomial distribution, respectively. In this study I developed and presented an algorithm and Java program to generate spatial distribution patterns for cell automata uses (Qi et al., 2011).

### 2 Algorithms

#### 2.1 Spatial distribution patterns

Supposedly there are three types of spatial distribution patterns, binomial distribution, Poisson distribution and negative binomial distribution (Krebs, 1999; Zhang, 2007). They represent the uniform distribution, random distribution and aggregated distribution, respectively. Recursive formulae for calculating probability of binomial distribution are:

$$p_r = q^n, \quad r=0;$$

$$p_r = p(n-r+1)p_{r-1}/(r q), \quad r=1, 2, \dots, n,$$

where,  $p$ : probability of a individual occurred in a cell,  $p = \sum i f_i / (n \sum f_i)$ ,  $q = 1 - p$ ,  $i = 0, 1, 2, \dots, n$ ;  $n$  is the maximum possible number of individuals in a cell,  $f_i$  is the number of cells with  $i$  individuals,  $i = 0, 1, 2, \dots, n$ .

Recursive formulae for calculating probability of Poisson distribution are:

$$p_r = e^{-m}, \quad r=0;$$

$$p_r = m p_{r-1}/r, \quad r>0,$$

where,  $m$ : mean of the number of individuals in a cell.

In the negative binomial distribution, the number of clusters per cell is Poisson distribution, the mean number of clusters is  $k \times \ln Q$ , and the number of individuals per cluster is the logarithmic distribution. The recursive formulae are:

$$p_r = (1-1/Q)^r / (r / \ln(1/Q)), \quad r > 0,$$

where,

$$Q = 1 + m/k, \quad k = m^2 / (s^2 - m).$$

and,  $m$ : mean number of individuals,  $s^2$ : variance of the number of individuals.

## 2.2 Algorithm for generation of spatial distribution patterns

Given  $p_r$ ,  $r=0, 1, 2, \dots$ , are obtained from the formulae above, and there are  $N$  cells. Then the number of cells with  $r$  individuals inside the cell is  $\text{int}(N p_{r+a})$ ,  $r=0, 1, 2, \dots, M-1$ , and:

$$\sum \text{int}(N p_{r+a}) = N,$$

where the value of  $a$ ,  $0 < a < 1$ , makes  $M$  the limited number, and  $\text{int}$  means to get integer number. Coding all cells with the ID number,  $1, 2, \dots, N$ . Similar to the Monte Carlo simulation used by Ferrarini (2011), generate a number between 1 to  $N$  using random number generator, e.g.,  $V_b$ , and let  $Q_1 = V_b$ . Remove  $V_b$ , and re-coding the remaining  $N-1$  numbers, and generate a number between 1 to  $N$ , e.g.,  $V_a$ . Repeat this process until  $Q_N = V_q$ . As a result, the number of individuals in cells  $Q_1, Q_2, \dots, Q_{\text{int}(p_0 N+a)}$  are zeros, and the number of individuals in cells  $Q_{\sum \text{int}(p_r N+a)+1}, \dots, Q_{\sum \text{int}(p_r N+a)}$  are  $i$ ,  $i=1, 2, \dots, M-1$ , where for the first summation term,  $r=0, 1, 2, \dots, i-1$ , and for the last summation term,  $r=0, 1, 2, \dots, i$ . For negative binomial distribution, use the above algorithm in each cell. An initial spatial distribution is thus generated.

## 3 Java Implementation

The algorithm above is implemented by a Java program (based on JDK 1.1.8) (Zhang, 2011), in which seven classes and an HTML file is included (<http://www.iaees.org/publications/software/index.asp>). The class, BioDistriProducer, performs all calculation, while the other class is called together to complete the entire task. The user will be asked to choose the type of spatial distribution patterns, to enter the required number of spatial distribution patterns, the number of cells along the  $x$ -axis and  $y$ -axis (Fig. 1). The codes for generating the map of spatial distribution, `ran()`, and for calculating spatial distribution, `subpr()`, are listed as follows:

```
public int[] ran() {
    ep=0.5;
    int s2;
    s2=(int)Math.round(p[0]*na+ep);
    r=1;
    s2+=subpr();
    while (s2!=na) {
        if (s2<na) {r++;
        if (r>n) {ep+=0.5;
```

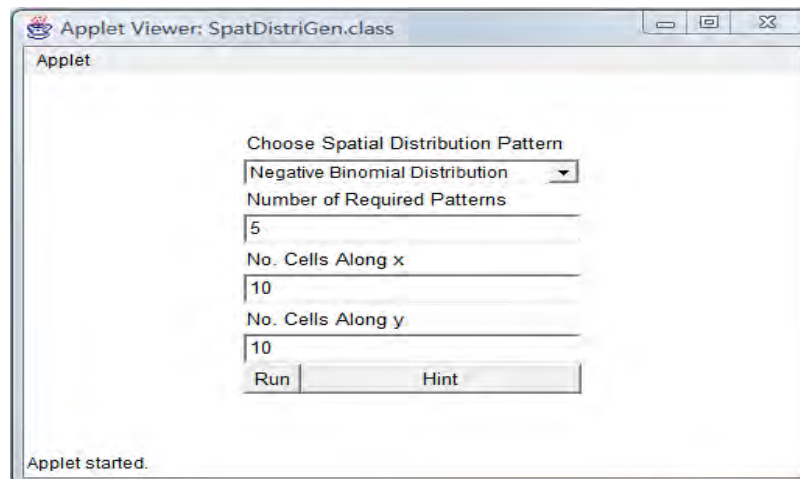
```

s2=(int)Math.round(p[0]*na+ep);
r=1;
s2+=subpr(); }
else if (r<=n) s2+=subpr(); }
else if (s2>na) {ep-=0.05;
s2=(int)Math.round(p[0]*na+ep);
r=1;
s2+=subpr(); } }
for(i=0;i<=r;i++) p[i]=Math.round(p[i]*na+ep);
s2=0;
for(i=1;i<=na;i++) x[i]=i;
for(i=1;i<=na;i++) {
if (i==na) break;
q[i]=(int)Math.round(Math.random()*(na-s2)+0.5);
h[i]=x[q[i]];
for(j=q[i]+1;j<=na-s2;j++) x[j-1]=x[j];
s2++; }
h[na]=x[1];
s2=1;
for(j=0;j<=r;j++) {
if (p[j]==0) continue;
for(int k1=0;k1<=p[j]-1;k1++) {x[h[s2]]=j; s2++; } }
test: for(i=1;i<=na;i++) {
for(j=1;j<=na;j++) if (h[j]==i) continue test;
x[i]=x[0]; break; }
return x; }

public int subpr() {
if (lm==1) p[r]=m/r*p[r-1];
else if (lm==2) p[r]=(nm-r+1)*p1*p[r-1]/(r*q1);
else if (lm==3) p[r]=(k+r-1)*m/k/(r*q1)*p[r-1];
return (int)Math.round(p[r]*na+ep); }

```

Further, a window will pop up to wait for entering parameters. Enter the mean number of individuals in a cell if Poisson distribution pattern was chosen. If binomial distribution pattern was chosen, enter the probability of an individual occurred in a cell, and the maximum possible number of individuals in a cell. For negative binomial distribution, enter the mean number of individuals in a cell and its variance.



**Fig. 1** An interface for the program

### 4 Application

Suppose the cell automata have 100 (10×10) cells. The mean number of individuals is 100 for Poisson distribution and mean number of individuals is 200 and variance is 20000 for negative binomial distribution (Fig. 2). A Poisson distribution pattern and a negative binomial distribution pattern, generated by SpatDistriGen, are as follows (and Fig. 2):

Poisson distri.									
97	96	103	101	90	102	103	84	109	92
93	94	97	106	101	78	107	98	104	90
106	95	94	79	108	89	88	85	92	97
99	101	100	98	94	100	95	89	91	87
110	102	103	110	74	101	96	93	103	104
109	106	85	105	77	97	102	92	81	76
83	100	93	96	99	105	86	94	100	98
82	91	102	98	95	80	106	104	109	93
99	107	107	86	89	84	75	91	104	99
110	90	96	108	87	105	88	95	108	105

Nega. bino. distri.									
13	44	70	76	11	79	33	38	68	52
4	40	27	31	16	84	14	86	60	17
74	36	97	25	48	21	57	23	34	41
45	39	6	35	90	63	59	12	55	83
65	99	7	19	58	89	43	85	28	30
80	72	20	32	82	5	8	51	29	77
56	2	61	50	96	49	10	22	46	26
66	37	3	69	18	9	88	62	64	93
47	87	95	92	71	24	98	54	53	75
94	67	73	0	91	78	1	42	15	81

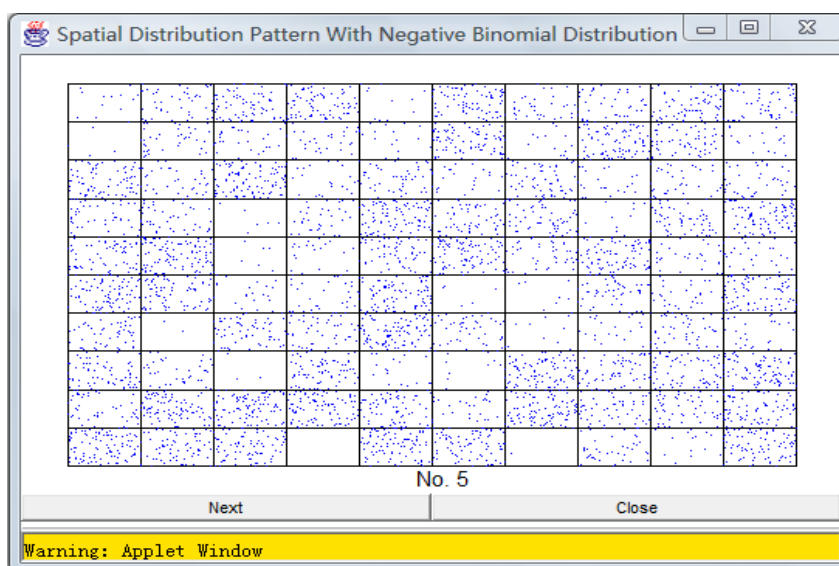


Fig. 2 Graphic output of initial spatial distribution.

The aggregation index,  $m^*/m$ , of Poisson distribution above is 1.009 and of negative binomial distribution is 1.3437. Based on the criteria for aggregation determination, the Poisson and negative binomial distributions above are random and aggregated distributions respectively, which is exactly correct.

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