Original Research Paper

Uniform Twister Plane Generator

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Corresponding Author: Yulian A. Menyaev Winthrop P. Rockefeller Cancer Institute, University of Arkansas for Medical Sciences, Little Rock, AR, USA Email: yamenyaev@uams.edu **Abstract:** Random plane generators may use various types of the random number algorithms to create multidimensional planes. At the same time, the discrete Descartes random planes have to be uniform. The matter is that using the concept of the uncontrolled random generation may lead to a result of weak quality due to initial sequences having either insufficient uniformity or skipping of the random numbers. This article offers a new approach for creating the absolute twisting uniform two-dimensional Descartes planes based on a model of complete twisting sequences of uniform random variables without repetitions or skipping. The simulation analyses confirm that the resulted random planes have an absolute uniformity. Moreover, combining the parameters of the original complete uniform sequences allows a significant increase in the number of created planes without using additional random access memory.

Keywords: Pseudorandom Number Generator, Stochastic Sequences, Congruential Numbers, Twister Generator, Random Plane, Random Field

Introduction

In our previous studies (Deon and Menyaev, 2016a; 2016b; 2017) there were proposed several pseudorandom number generators, particularly *nsDeonYuliTwist32D*, which offers a technique of using no congruential twisting array. This generator allows the creation of absolutely complete twister uniform sequences having various lengths.

The direction of Random Plane (RP) Generators (RPG) employs a stochastic process at the time of creating the points distributed on N-dimensional plane. Here we consider a two-Dimensional (2D) plane only. Other discrete-dimensional planes have the same initial properties. Each coordinate of RP-generated points may belong to its own Random Field (RF). An analysis of the last sources sums up the following selected types of random fields: Conditional RF (Quattoni et al., 2004; Sutton and McCallum, 2012), Markov RF (Sarawagi and Cohen, 2004; Bekkerman et al., 2006), Gaussian RF (Rimstad and Omre, 2014), uniformed RF (Xiao, 2010) and others (Qi et al., 2004; Dachian and Nahapetian, 2009). In the application areas the RPGs are often applied in graphical images (Kumar and Hebert, 2003), phone systems (Sung and Jurafsky, 2009), advertising applications, etc. Next, the RPGs are actively used in fundamental studies, starting from 2D theoretical modeling (Gnedenko, 1998; Feller, 2008), Monte Carlo plane simulation (Newman and Barkema, 1996; Spanos and

Zeldin, 1998), factorial development (Kim and Zabih, 2002), realizations for training systems (Sha and Pereira, 2003), etc. and biomedical engineering (Menyaev and Zharov, 2005; 2006a; 2006b; Menyaev *et al.*, 2013; 2016; Koonce *et al.*, 2017).

The principles of all these studies are based on the conception of random planes, in which the Descartes plane features have to satisfy the following properties: (1) The generation process has to provide the uniqueness (i.e., no repetitions) of each point on the plane and (2) the generation process has to keep the completeness (i.e., without skipping) for all created points. These properties should be considered as a 'natural filter' for choosing the random number generator.

Let's consider two of them in brief. If the generation uses the twister generator MT19937 (Matsumoto and Nishimura, 1998; Matsumoto *et al.*, 2006; 2007; Saito and Matsumoto, 2008), then the result of this attempt is very discussable since this generator in DieHard Tests (Berger and Zorn, 2006; Novark and Berger, 2010; Alani, 2010) demonstrates a uniqueness level of 0.7, which is equivalent to the level of repeatability 1-0.7 = 0.3. On the other hand, we may use the twister generator *nsDeonYuliTwist32D* (Deon and Menyaev, 2017), which is guaranteed to create the complete uniform twisting sequences of an arbitrary size having no repetitions and skipping of elements. Now the question here is: Would it be possible to observe the Descartes properties in the current particular task?



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Let's consider this issue more in detail. For this, let us use the aforementioned twister generator, which is capable of creating complete uniform twisting sequences of arbitrary size. Below is the program code for modeling a grid of the discrete Descartes plane $U \times V$. The program algorithm generates the integer random numbers independently along the U and V independent axes. Matrix A is the discrete plane indicator; its cells $a[u, v] \in A$ with indices u and v correspond to the coordinates of discrete points $\langle u, v \rangle \in U \times V$. The value of cell $a[u, v] \in A$ indicates the quantity of attempts to create independently the corresponding point $\langle u, v \rangle$ on the generated plane. Without loss of generality and in order to visualize the result, we assign the amount of discrete coordinates by given sets $U = V = \{0, 1, 2, 3, 4, 5, ...\}$ 6, 7}, which in the binary representation corresponds to the length w = 3 bits for each coordinate. According to the previous studies (Deon and Menyaev, 2016b; 2017), let's choose a twister generator nsDeonYuliTwist32D, which operates on the basis of congruential model $x_{i+1} = (ax_i + c)$ & maskW with constants a = 5, c = 1. The initial values of the twisting sequences are taken here as $x_0 = 1 \in U$ and x_0 $= 4 \in V$. Any other choice of parameters for generation is possible; the essence of the obtained results will not be changed. Program names P040101 and cP040101 are taken by chance. The chosen programming language is C# available in Microsoft Visual Studio. The use of other dialects of the older C versions (i.e., Win32) or C++ (CLR) provides the same results.

using nsDeonYuliTwist32D; // twister uniform generator namespace P040101

{ class cP040101	
<pre>{ static void Main(string[] and</pre>	·gs)
$\{ uint w = 3; \}$	// number bit length
cDeonYuliTwist32D GU	
new cDe	conYuliTwist32D();
GU.x0 = 1;	// U sequence beginning
GU.w = w;	// number bit length
GU.Start();	// GU generator starts
cDeonYuliTwist32D GV	r =
new cD	eonYuliTwist32D();
GV.x0 = 4;	// V sequence beginning
GV.w = w;	// number bit length
GV.Start();	// GV generator starts
	U and V sequences length
Console.WriteLine("w =	$\{0\}$ N = $\{1\}$ ", w, N);
uint[] U = new uint[N];	// U sequence
uint[] V = new uint[N];	// V sequence
int[,] A = new int[N,N];	// result matrix
for (int $i = 0; i < N; i++$)	
for (int $j = 0; j < N; j+-$	+) $A[i, j] = 0;$
for (int $i = 0; i < N; i++$)	// one of the axis
{ Console.Write(" $i = \{0\}$,3} ", i);
for (int $j = 0; j < N; j+$	+) // another axis

```
{ uint u = GU.Next();
                                // u random number
      U[i] = u;
                              // random U sequence
      uint v = GV.Next();
                                // v random number
                              // random V sequence
      V[j] = v;
                  // <u,v> point generation counter
      A[u, v] ++;
   }
   Console.Write("U = ");
   for (int m = 0; m < N; m++)
     Console.Write("{0,4}", U[m]);
   Console.WriteLine();
   Console.Write("
                        | V = "):
   for (int m = 0; m < N; m++)
     Console.Write("{0,4}", V[m]);
   Console.WriteLine();
  }
 Console.WriteLine("Matrix A");
  for (int i = 0; i < N; i++)
  {for (int j = 0; j < N; j++)
     Console.Write("{0,4}", A[i, j]);
   Console.WriteLine();
 Console.ReadKey();
                                    // result viewing
}
```

After this code execution the listing below appears:

$\mathbf{W} =$	3 N	V = 8								
i =	0	U =	1	6	7	4	5	2	3	0
		$\mathbf{V} =$	4	5	2	3	0	1	6	7
i =	1	U =	3	5	7	1	2	4	6	0
		$\mathbf{V} =$	1	2	4	6	0	3	5	7
i =	2	U =	7	3	6	2	5	1	4	0
		V =	2	5	1	4	0	7	3	6
i =	3	U =	6	7	4	5	2	3	0	1
			5	2	3	0	1	6	7	4
i =	4	U =	5	7	1	2	4	6	0	3
		$\mathbf{V} =$	2	4	6	0	3	5	7	1
i =	5	U =	3	6	2	5	1	4	0	7
		$\mathbf{V} =$	5	1	4	0	7	3	6	2
i =	6	U =	7	4	5	2	3	0	1	6
		V =	2	3	0	1	6	7	4	5
i =	7	U =	7	1	2	4	6	0	3	5
		$\mathbf{V} =$	4	6	0	3	5	7	1	2

Matrix A

} }

0	0	0	0	0	0	2	6
0	0	0	0	3	0	3	2
3	3	0	0	2	0	0	0
0	3	0	0	0	2	3	0
0	0	0	8	0	0	0	0
5	0	3	0	0	0	0	0
0	2	0	0	0	6	0	0
0	0	5	0	3	0	0	0

In matrix A the cells with values that differed from 1, show that the independent generation of coordinates of the points on the plane doesn't ensure uniform distribution of random points $\langle u, v \rangle$. Some points are missed (values 0) and others are present several times (values > 1).

So, the aim of this article is to find a solution for the generation of uniform discrete twisting planes, which possess the Descartes property of a single presence of the random points in nodes of the discretization grid.

Theory

One of the options to represent the discrete Descartes plane is an enumeration of all points in the grid nodes, formed from the values of discretization on the corresponding axes. If the location of axes is independent, the grid has a rectangular view. Moreover, if discretization for both axes is the same and uniform, the grid has a square view.

Let's assume that the square grid includes N points of discretization along each axis. Thus, the total number of grid points is $N \times N = N^2$. To set these points in a random way, an algorithm is required that can provide a random move from one point to another.

Now would be a good time to point out and emphasize the following: *The plane is random only if moving from one point to another, while creating the plane, utilizes the stochastic process.*

In this case, the requirement of Descartes axes, which prescribes a unique representation of each point, has to be kept. The unambiguity is determined by the discretization of the axes. Uniqueness is provided by an appropriate procedure, which does not allow entering each point of the grid twice or more times. The skipping of vertices of the grid isn't allowed either. In other words, each point is presented once during generation of all the grid points. In this case, the total enumeration of the points is N^2 . Following this way, such a grid on the Descartes plane is called *uniform* and a random Descartes plane, which contains uniform grid, is called uniform Descartes RP.

There are many ways to specify the points on the grid. Let's name a few of them:

- Rectangular left or right filling of the grid, when one of the axes is selected and at each location of discretization of this axis, the points along the discrete points of the other axis are placed on the grid
- Rectangular top or bottom filling of the grid under the same conditions
- Diagonal filling of the grid under the same conditions
- The secondary indexing of the discrete points along the Descartes axes

This is not a whole list of possible techniques. The options to choose aren't limited and can be organized by the designer in any possible way. Note that items (1) - (3) create ordinary Descartes planes and item (4) allows obtaining the random Descartes planes, if the secondary index is a result of the stochastic process.

In this article, an option of secondary indexing of the discrete marks on the Descartes axes is adopted. Let's demonstrate this by an example, in which the congruential generation of random numbers $x_{i+1} = (ax_i + ax_i)$ c)&maskW is used as secondary indexing base. In order to visualize the results, we take the complete uniform sequences of random numbers $x \in \{0, 1, 2, 3\} = \{00_2, \dots, 0\}$ 01_2 , 10_2 , 11_2 } having length w = 2 bits. In this case, each complete sequence contains $N = 2^w = 2^2 = 4$ elements. Without loss of generality, let's assume that $a = 1 \in [\overline{1, N-1}]$ and $c = 3 \in [\overline{1, N-1}]$. In total, four congruential sequences are possible: <0, 3, 2, 1>, <1, 0, 3, 2>, <2, 1, 0, 3>, <3, 2, 1, 0>. These sequences allow creating various random tracks on uniform Descartes RP.

If random value $x_0 = 1$ is chosen as an initial value, then the designated generator *GU* creates the sequence *U* = <1, 0, 3, 2>. From this it follows that the initial random vertex will be located on the vertical part of the grid with horizontal discrete mark 1 along the *U* axis (Fig. 1).

If the second independent generator, which is designated as GV, uses the initial random value $x_0 = 3$, then sequence $V = \langle 3, 2, 1, 0 \rangle$ is created. From this it's obvious that the second coordinate has the value of 1 for the initial random point $\langle 1, 3 \rangle$. The next vertex has coordinates $\langle 0, 2 \rangle$. Both obtained vertices are connected by an arc, forming the beginning of the random track. Then, vertex $\langle 3, 1 \rangle$ will be placed on this track. Finally, the vertex with coordinates $\langle 2, 0 \rangle$ completes the random track is shown in Fig. 1.

Regarding the random sequence of secondary indices <1, 0, 3, 2> along the *U* axis, four sequences along the *V* axis are possible:

1) U = <1, 0, 3, 2> V = <3, 2, 1, 0>2) U = <1, 0, 3, 2> V = <2, 1, 0, 3>3) U = <1, 0, 3, 2> V = <1, 0, 3, 2> V = <1, 0, 3, 2>4) U = <1, 0, 3, 2>V = <0, 3, 2, 1>

These sequences V can be interpreted as the left circular shift (Deon and Menyaev, 2016b) of the original sequence $\langle 3, 2, 1, 0 \rangle$. Figure 2 shows four tracks of an interaction of the pairs of sequences on the U and V axes. These tracks provide an exact one-time generation of each vertex on a grid of the Descartes RP.

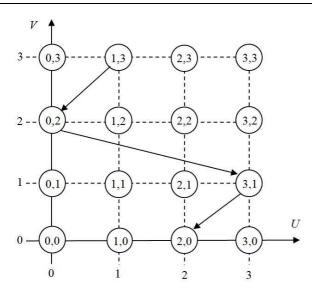


Fig. 1: The initial random track

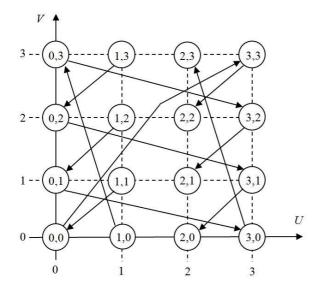


Fig. 2: All the tracks of the random sequences U and V

The program code for this task is presented below, in which all the vertices are generated on a grid of the Descartes RP. Random numbers have the length of w = 3 bits. In each sequences U and V there are $N = 2^w = 2^3 = 8$ random numbers with $x \in [0, N-1] = \{0, 1, 2, 3, 4, 5, 6, 7\}$. Both sequences U and V are given by a congruential model $x_{i+1} = (ax_i + c)$ &maskW with coefficients a = 5 and c = 1. The first sequence begins with value $x_0 = 1$ and has the form U = <1, 6, 7, 4, 5, 2, 3, 0>. The second sequence begins with value $x_0 = 4$ and has the form V = <4, 5, 2, 3, 0, 1, 6, 7>. Each cell of matrix A corresponds to one vertex on a grid of RP. The value of cell $a \in A$ shows how many times the corresponding vertex is generated. Program names P040201 and cP040201 are selected by chance.

namespace P040201

{ class cP040201

{ static void Main(string[] args) { uint w = 3; // number bit length int $N = 1 \ll (int)w$; // U and V sequences length Console.WriteLine(" $w = \{0\}$ N = $\{1\}$ ", w, N); uint[] U = new uint[8] {1,6,7,4,5,2,3,0}; uint[] V = new uint[8] $\{4,5,2,3,0,1,6,7\};$ int[,] A = new int[N, N];// result matrix for (int i = 0; i < N; i++) for (int j = 0; j < N; j++) A[i, j] = 0; for (int i = 0; i < N; i++) // one of the axis { Console.Write(" $i = \{0,3\} \mid ", i$); for (int j = 0; j < N; j++) // another axis A[U[j], V[j]]++; // <u,v> generation counter Console.Write("U = "); for (int m = 0; m < N; m++) Console.Write("{0,4}", U[m]); Console.WriteLine(); V = "): Console.Write(" for (int m = 0; m < N; m++) Console.Write("{0,4}", V[m]); Console.WriteLine(); // V shift beginning uint r = V[0]; for (int m = 1; m < N; m++) V[m - 1] = V[m]; V[N-1] = r;Console.WriteLine("Matrix A"); for (int i = 0; i < N; i++) { for (int j = 0; j < N; j++)Console.Write("{0,4}", A[i, j]); Console.WriteLine(); Console.ReadKey(); // result viewing } }

After this code execution the listing below appears:

Matrix A

1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	
1	1	1	1	1	1	1	1	

The values of 1 in the cells of matrix A show that each cell was updated once. This result reflects the single initialization of the corresponding vertices on a grid of uniform Descartes RP.

Similar tests of creating uniform planes with an arbitrary bit length w of the random numbers confirm a uniformity of the received RPs. Replacement of the congruential sequences by the different corresponding twisting sequences shows a similar result when generating the twisting RPs. Thus, the complete uniform twisting sequences can ensure a creation of the complete uniform twisting RPs. Now we may proceed to the generator designing.

Construction and Results

Program P040201, described in the previous section, utilizes two arrays U and V as initial congruential sequences. Such arrays can be created using the twister generator nsDeonYuliTwist28DA (Deon and Menyaev, 2016b). However, this solution will not be perfect for the case of large planes because it requires a lot of available Random Access Memory (RAM). For many reasons it may be unavailable on a computer. This limitation can be generator overcome by using the twister nsDeonYuliTwist32D (Deon and Menyaev, 2017), which doesn't use arrays of the twisting sequences. However, a more satisfactory solution is the aforementioned program P040201, which implies the repeated generation of U and V, whereas nsDeonYuliTwist32D doesn't provide this. Thus, we come to the inference that before performing a generation of the twisting planes, it is necessary to have tools for the trivial operations with the twisting sequences.

Simple Twister Generator

Class *cDeonYuliSTwist32D*, which is presented below, includes in its name a letter *S* indicating the meaning of the word 'simple'. This class provides elementary operations with the twisting sequences and it does not use any arrays. The prototype of class *cDeonYuliSTwist32D* is class *cDeonYuliTwist32D*. They are different in the issue that automatic setting of the congruential parameters *a* and *c* is excluded in class *cDeonYuliSTwist32D*. An example of using this class is presented in this section later, in the description of the program code *P040301*.

namespace nsDeonYuliSTwist32D

{ class cDeonYuliSTwist32D)
$\{ public uint w = 16U; \}$	// number bit length
public uint $N1 = 0U$;	// max-number
public uint $x0 = 1U$;	// sequence beginning
uint $xB = 1U;$	// twister beginning
public uint $xG = 0U;$	// created random number
uint $xL = 0U$, $xR = 1U$;	// pair numbers
public uint $a = 5U$;	// congruential constant a
public uint $c = 1U$;	// congruential constant c
public uint mask $W = 0U$;	// number mask
public uint mask $U = 0U;$	// elder bit mask
public uint mask $T = 0U;$	// twister bits
public uint $nW = 0U;$	// pair twister number in w

public cDeonYuliSTwist32D()
{ N1 = 0xFFFFFFFF >> (32 - (int)w);//max-number
}

,
<pre>public void StartCong(uint sxB) { xB = x0; uint sxBe = sxB & maskW; for (int i = 0; i < sxBe; i++) xB = (a * xB + c) & maskW;// shifted beginning xR = xB;</pre>
}
<pre>public uint NextCong() { xL = xR;</pre>
<pre>public void RepeatCong() { xR = xB; // repeat track }</pre>
<pre>// public void ShiftCong() { xB = (a * xB + c) & maskW; // shifted beginning</pre>
<pre>// public void StartTwist(uint snW) { nW = (uint)snW;</pre>
//

Aleksei F. Deon and Yulian A. Menyaev / Journal of Computer Science 2018, 14 (2): 260.272 DOI: 10.3844/jcssp.2018.260.272

xR = (a * xL + c) & maskW; // end of xL,xR pair // twister of pair return xG; } //____ public void RepeatTwist() $\{ xL = xB;$ // for twister beginning xR = (a * xL + c) & maskW; // end of xL, xR pair} //----public void ShiftTwist() $\{xB = (a * xB + c) \& maskW; // shifted beginning \}$ xL = xB;// for twister beginning xR = (a * xL + c) & maskW; // end of xL,xR pair} //----public void Start() { N1 = 0xFFFFFFFF >> (32- (int)w);//max-number maskW = 0xFFFFFFF >> (32- (int)w); //n. mask maskU = 1U << ((int)w - 1);// elder bit mask maskT = maskU;// first twister bit // a-value DeonYuli PlusA(); DeonYuli SetC(); // c-value x0 &= maskW;// sequence beginning StartCong(0); // for congruential generation } //____ public void TimeStart() { x0 = (uint)DateTime.Now.Millisecond;// millisecs // generator starts Start(); } //----public void SetW(uint sw) $\{ w = (uint)Math.Abs(sw); // number bit length \}$ DeonYuli_SetW(); } //____ public void SetW(int sw) $\{ w = (uint)Math.Abs(sw); // number bit length \}$ DeonYuli SetW(); } _____ public void DeonYuli SetW() // min-length // max-length { if (w < 3U) w = 3U;else if (w > 32U) w = 32U; N1 = 0xFFFFFFFF >> (32-(int)w);//max-numberx0 = N1 / 7U;// sequence beginning } _____ //_____ public void SetA(double sa) { double ad = Math.Abs(sa); if (ad > 1.0) ad = 1.0;a = (uint)(N1 * ad); // related set of a } //----public void SetA(uint sa) { a = (uint)Math.Abs(sa); if (a < 1) a = 1; // min-value

if (a > N1) a = N1;// max-value } _____ void DeonYuli PlusA() { if (a < 1U) {a = 1U; return; } uint z = a; // bottom edge for a for (uint i = 0U: i < 3U: i++) if (a % 4U != 0U) a--;// random condition else break; // true value for constant a a++: if (a < z) a += 4U; // on right from bottom edge if $(a \ge N1 - 1) a = 4U$; // on left from top edge } //____ public void SetC(double sc) { double cd = Math.Abs(sc); if (cd > 1.0) cd = 1.0;c = (uint)(N1 * cd);// related set of c } public void SetC(uint sc) $\{ c = (uint)Math.Abs(sc); \}$ if (c < 1) a = 1;// min-value if (c > N1) c = N1; // max-value } //___ void DeonYuli SetC() { if (c % 2U = 0U)c + 1; // only odd c if (c > N1) c = N1;// max-value } _____ public void SetX0(double sx) { double xd = Math.Abs(sx); if (xd > 1.0) xd = 1.0; x0 = (uint)(N1 * xd);// sequence beginning } _____ public void SetX0(int sx) $\{ x0 = (uint)sx; \}$ // sequence beginning } //= } }

To verify the correct utilization of the presented generator *nsDeonYuliSTwist32D*, let's use the program code *P040301*, in which the sequences of all twisters of the random numbers having length w = 3 bits are generated. The total quantity of sequences is $w \cdot N = w \cdot 2^w = 3 \cdot 2^3 = 24$. Program names *P040301* and *cP040301* are taken by chance.

using nsDeonYuliSTwist32D; // s-twister uni-generator namespace P040301

{ class cP040301

{ static void Main(string[] args)

{ cDeonYuliSTwist32D ST =

new cDeonYuliSTwist32D(); ST.SetW(3); // number bit length // congruential constant a ST.SetA(5);// congruential constant c ST.SetC(1);// sequence beginning ST.SetX0(1); // generator starts ST.Start(); int w = (int)ST.w;// random bit length int N = (int)ST.N1 + 1;// sequence length Console. WriteLine(" $w = \{0\}$ N = $\{1\}$ ", w, N); Console. WriteLine(" $a = \{0\}$ $c = \{1\}$ $x0 = \{2\}$ ". ST.a, ST.c, ST.x0); int k = 0: for (int m = 0; $m \le ST.N1$; m++) { Console.Write(" $k = \{0,3\}$ | Cong = ", k++); for (int n = 0; $n \le ST.N1$; n++) { uint z = ST.NextCong(); //congruential number Console.Write(" $\{0,4\}$ ", z); Console.WriteLine(); for (int nW = 1; nW < ST.w; nW++) { Console.Write(" $k = \{0,3\}$ | Twist $\{1\} =$ ", k++, nW; ST.StartTwist((uint)nW); // twist beginning nW for (int n = 0; $n \le ST.N1$; n++) { uint z = ST.NextTwist(); // twister number Console.Write("{0,4}", z); Console.WriteLine(); ST.RepeatTwist(); ST.ShiftCong(); // congruential shift Console.ReadKey(); // result viewing } }

After the execution of *P040301*, the following listing appears on the monitor:

```
w = 3 N = 8
a = 5 \ c = 1 \ x0 = 1
k = 0 | Cong =
                 1 6 7 4 5 2 3 0
k = 1 | Twist 1 = 3 5 7 1 2 4 6 0
k = 2 | Twist 2 = 7 3 6 2 5 1 4 0
k = 3 | Cong = 6 7 4 5 2 3 0 1
k = 4 | Twist 1 = 5 7 1 2 4 6 0 3
k = 5 | Twist 2 = 3 6 2 5 1 4 0 7
       Cong = 7 4 5 2 3 0 1 6
k = 6 |
       Twist 1 = 7 1 2 4 6 0 3 5
k = 7 |
k = 8 | Twist 2 = 6 2 5 1 4 0 7 3
k = 9 | Cong =
                 4 5 2 3 0 1 6 7
k = 10 | Twist 1 = 1 2 4 6 0 3 5 7
k = 11 | Twist 2 = 2 5 1 4 0 7 3 6
k = 12 | Cong =
                 5 2 3 0 1 6 7 4
k = 13 | Twist 1 = 2 4 6 0 3 5 7 1
```

}

k = 15 k = 16 k = 17 k = 18 k = 19 k = 20 k = 21	Cong = Twist 1 = Twist 2 = Cong = Twist 1 = Twist 2 = Cong =	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	
k = 22	Twist 1 =	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	

These congruential twisting sequences coincide with the result of tests of our previously developed twister generator *nsDeonYuliTwist28DA* utilizing a technique of the congruential twisting array.

Running the program P040301 with other values of the bit length w of the random numbers confirms a completeness of the generated sequences. This is sufficient to ensure the development of programs for generating the Descartes uniform twisting RPs.

Twister Generator of Uniform Planes

When constructing a complete generator of uniform twisting planes, let's use two uniform complete generators GU and GV of the aforementioned instrumental class cDeonYuliSTwist32D. It allows organizing the next class nsDeonYuliPlaneTwist32D for the generation of all points on a grid of the twisting plane. By default, the initial track takes a diagonal of the grid points from the left-bottom position to the right-top one, although this may be changed by setting the independent beginnings for the internal generators GU and GV. An example use of class cDeonYuliPlaneTwist32D is presented in this section later, located in the description of the program code P040302.

 $using \ nsDeonYuliSTwist32D; \ \ // \ s-twister \ uni-generator \\ namespace \ nsDeonYuliPlaneTwist32D$

{ class cDeonYuliPlaneTwist32D

public uint w = 16; // uniform number bit length £ public uint N1 = 0; // max-number in track public uint a = 5U; // congruential constant a public uint c = 1U; // congruential constant c public unit x0 = 1U; // constant of track beginning public cDeonYuliSTwist32D GU; // generator 1 public cDeonYuliSTwist32D GV; // generator 2 public unt nWU = 0; // twister shift number in GU public uint nRU = 0; // ring shift number in GU public uint nWV = 0; // twister shift number in GV public uint nRV = 0; // ring shift number in GV public unit nG = 0; // elements in GU and GV tracks -----

public cDeonYuliPlaneTwist32D()

```
{ GU = new cDeonYuliSTwist32D(); // generator 1
  GV = new cDeonYuliSTwist32D(); // generator 2
}
```

```
public void SetW(int sw)
```

Aleksei F. Deon and Yulian A. Menyaev / Journal of Computer Science 2018, 14 (2): 260.272 DOI: 10.3844/jcssp.2018.260.272

 $\{ w = (uint)sw; \}$ // random number bit length DeonYuli SetWN1ACX(); } public void SetW(uint sw) $\{ w = sw; \}$ // random number bit length DeonYuli SetWN1ACX(); //_ void DeonYuli SetWN1ACX() { N1 = 0xFFFFFFF >> (32- (int)w);//max-number a = (uint)((double)N1*0.39);//congruential const ac = a / 2;// congruential const c x0 = N1 / 2U;// constant of track beginning } //_ void DeonYuli_SetN1ACX() // random number bit length in GU $\{ GU.w = w; \}$ GU.a = a;// congruential constant a for GU GU.c = c;// congruential constant c for GU GU.x0 = x0;// sequence beginning in GU GV.w = w;// random number bit length in GV GV.a = a;// congruential constant a for GV GV.c = c;// congruential constant c for GV GV.x0 = x0;// sequential beginning in GV } //_ public void Start() DeonYuli SetN1ACX(); // congruential constants { GU.Start(); // GU generator starts GV.Start(); // GV generator starts a = GU.a;// congruential constant a c = GU.c;// congruential constant c x0 = GU.x0; // beginning of U and V sequences nWU = 0; // twister shift number in GU generator nRU = 0;// ring shift number in GU generator nWV = 0; // twister shift number in GV generator nRV = 0;// ring shift number in GV generator nG = 0; // elements number in GU and GV tracks GU.StartCong(0);//for congruential GU generation GV.StartCong(0);//for congruential GV generation } //_ public void Next(ref uint u, ref uint v) if (nWU == 0) GU.NextCong(); £ else GU.NextTwist(); if (nWV == 0) GV.NextCong(); else GV.NextTwist(); u = GU.xG;// <u,v> point coordinates v = GV.xG;if (nG < N1) {nG++; return; } // inside track nG = 0;// regular track beginning if (DeonYuli RingCongGV()) return;// inside GV if (DeonYuli RingTwistGV()) return;// inside GV if (DeonYuli RingCongGU()) return;// inside GU DeonYuli RingTwistGU(); // inside GU

//	}
	<pre>bool DeonYuli_RingCongGV() { if (nWV != 0) return false; // no congruential ring if (nWU == 0) GU.RepeatCong(); else GU.RepeatTwist(); if (nRV < N1)//congruent. ring opportunity in GV { GV.ShiftCong(); // congruential shift in GV nRV++; // next ring number return true; // inside congruential track in GV } pRV = 0; // first ring in CV</pre>
	nRV = 0;// first ring in GVnWV = 1;// first twister in GVGV.StartTwist(nWV);// twister starts in GVreturn true;// inside GV
,,	}
//	<pre>bool DeonYuli_RingTwistGV() { if (nWV == 0) return false; // no twister ring if (nWU == 0) GU.RepeatCong(); else GU.RepeatTwist(); if (nRV < N1) // twister ring opportunity in GV { GV.ShiftTwist(); // twister shift in GV nRV++; // next ring number return true; // inside twister track in GV } nRV = 0; // new ring in GV if (nWV < w - 1) // continue twisters { nWV++; // next bit twister in GV GV.StartTwist(nWV); // twister starts in GV return true; // inside twister regime in GV return true; // inside twister op in GV return true; // new ring in GV return true; // inside twister regime in GV GV.StartTwist(nWV); // twister starts in GV return true; // inside twister regime in GV } nRV = 0; // new ring in GV gV.StartCong(nWV); // GV generator starts return false; // rings R2 and W2 are over in GV }</pre>
//	<pre>bool DeonYuli_RingCongGU() { if (nWU != 0) return false; // no congruential ring if (nRU < N1)//congruent. ring opportunity in GU { GU.ShiftCong(); // congruential shift in GU nRU++; // next ring number return true; // inside congruential track in GU } nRU = 0; // first ring in GU nWU = 1; // first twister in GU GU.StartTwist(nWU); // twister starts in GU return true; // inside GU }</pre>
//	<pre>bool DeonYuli_RingTwistGU() { if (nWU == 0) return false; // no twister ring if (nRU < N1) // twister ring opportunity in GU { GU.ShiftTwist(); // twister shift in GU nRU++; // next ring number</pre>

Aleksei F. Deon and Yulian A. Menyaev / Journal of Computer Science 2018, 14 (2): 260.272 DOI: 10.3844/jcssp.2018.260.272

return true; // inside twist track in GU } nRU = 0;// new ring in GU // continue twisters if (nWU < w - 1)// next bit twister in GU { nWU++; GU.StartTwist(nWU); // twister starts in GU // inside twister regime in GU return true: nRU = 0;// new ring in GU nWU = 0; // congruent. beginning (twister 0) in GU GU.StartCong(nWU); // GU generator beginning return false;//all planes created; common beginning } //= } }

To test an operation of the presented generator *nsDeonYuliPlaneTwist32D*, let's use the program code *P040302* shown below, in which the points of the initial twisting plane of the random numbers having length w = 3 bits are generated. By default, the initial track of generation is a diagonal of the grid points from the leftbottom position to the right-top one. The uniformity of points on a plane would be confirmed by matrix *A*, in which the cell values are the counters of generation of the corresponding points < u, v > on RP. Program names *P040302* and *cP040302* are taken by chance.

using nsDeonYuliPlaneTwist32D;//twist-plane generator namespace P040302

{ class cP040302
{ static void Main(string[] args)
 { cDeonYuliPlaneTwist32D TP =
 new cDeonYuliPlan

new cDeonYuliPlaneTwist32D(); int w = 3; // random number bit length int $N = 1 \ll w$; // track length TP.SetW(w); TP.Start(); // generator starts Console.WriteLine(" $w = \{0\}$ N = $\{1\}$ ", w, N); Console. WriteLine(" $a = \{0\}$ $c = \{1\}$ $x0 = \{2\}$ ", TP.a, TP.c, TP.x0); uint[] u = new uint[N]; // point u-coords on track uint[] v = new uint[N]; // point v-coords on track int[,] A = new int[N, N];// result matrix for (int i = 0; i < N; i++) for (int j = 0; j < N; j++) A[i, j] = 0; uint uu = 0; // point job coordinates on plane uint vv = 0: for (int i = 0; i < N; i++) // track values on plane { Console.Write(" $i = \{0,4\}$ ", i); Console.Write(" $nWU = \{0,3\}$ ", TP.nWU); Console.Write(" $nRU = \{0,3\}$ ", TP.nRU); Console.Write(" $nWV = \{0,3\}$ ", TP.nWV); Console.Write(" $nRV = \{0,3\}$ ", TP.nRV); Console.WriteLine();

for (int j = 0; j < N; j++) { TP.Next(ref uu, ref vv); // point on grid u[j] = uu;v[j] = vv;// generation counter for <u,v> A[uu, vv]++; 3 U = "); Console.Write(" for (int j = 0; j < N; j++) Console.Write("{0,4}", u[j]); Console.WriteLine(); V = "):Console.Write(" for (int j = 0; j < N; j++) Console.Write("{0,4}", v[j]); Console.WriteLine(); ł Console.WriteLine("Matrix A"); for (int i = 0; i < N; i++) { for (int j = 0; j < N; j++) Console.Write("{0,4}", A[i, j]); Console.WriteLine(); Console.ReadKey(); // result viewing }

After the execution of *P040302* code, the listing below appears. To reduce the listing size, we skipped some strings, which are indicated by a dashed line.

```
w = 3 N = 8
a = 5 c = 1 x0 = 3
i = 0 nWU = 0 nRU = 0 nWV = 0 nRV = 0
        U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
        V =
              3 0 1 6 7 4 5 2
    1 \text{ nWU} = 0 \text{ nRU} = 0 \text{ nWV} = 0 \text{ nRV} = 1
i =
        U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
        V = 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2 \ 3
    2 nWU = 0 nRU = 0 nWV = 0 nRV = 2
i =
        U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
        V = 1 \ 6 \ 7 \ 4 \ 5 \ 2 \ 3 \ 0
- - - - -
    7
       nWU = 0 nRU = 0 nWV = 0 nRV = 7
i =
        U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
        V = 2 3 0 1 6 7 4 5
Matrix A
1 1 1 1 1 1 1 1
1
  1
     1
        1
           1
              1
                 1 1
1
   1
      1
            1
         1
               1
                  1
1
   1
      1
         1
            1
               1
                  1
1
   1
      1
         1
            1
               1
                  1
   1
1
      1
         1
            1
               1
                  1
     1
        1
           1
1
   1
              1
                  1
                     1
   1
     1 1 1 1
1
                 1 1
```

In this listing, indicator nWU shows the number of a

}

twister on axis U and indicator nRU is pointed to the circular shift number on this axis U. The values nWU = 0 and nRU = 0 correspond to the congruential initial sequence on axis U. Similar values of indicators nWV and nRV show the circular shift of the initial sequence on axis V.

The listing of results for the singular matrix A confirms that each random point $\langle u, v \rangle$ is created once. The same results are valid for other $w \leq 32$. This corresponds to a concept of the Descartes uniform RP. Thus, generator *nsDeonYuliPlaneTwist32D* is ready for implementation.

Discussion

Twister generator *nsDeonYuliPlaneTwist32D* is capable creating a set of uniform twisting RPs for each pair of the congruential constants $a, c \in [\overline{1, N-1}] = [\overline{1, 2^{W} - 1}]$ in the sequences of random numbers having w bit length. All the initial values $x_0 \in [\overline{0, 2^{W} - 1}]$ are automatically presented in circles of the twisters (Deon and Menyaev, 2016b). However, the question is how many twisting planes could be obtained for each pair of parameters *a* and *c*?

In the previous section, it has been determined that when creating the initial RP $U \times V$, two random sequences U and V can be taken, which are created by the corresponding twisting generators. It is known that twister 0 is the initial uniform congruential sequence. Let's denote it as U_0 . Another sequence V_0 could be chosen arbitrarily but on the condition that it is also uniform. If the sizes or quantity of elements in the initial sequences are the same card $(U_0) = card (V_0)$, then it can be stated that among all the possible tracks of the random uniform plane there has to be a track that creates points on the second diagonal (from the left-bottom position to the right-top one) of the corresponding square discrete grid. This is somewhat reminiscent of the idea of a central abstract element and specifically in our case this second diagonal track is the central track of the discrete grid. In generator nsDeonYuliPlaneTwist32D this is exactly what is done, shown at the beginning of the listing of results for the previous program P0040302.

$U_0 =$	3	0	1	6	7	4	5	2
$V_0 =$	3	0	1	6	7	4	5	2

This is not the only solution because as the central track, one could take the main diagonal (from the left-top position to the right-bottom one) of a grid. But since the initial sequences U and V are random, the order of the vertex passage <3, 3>,<0, 0>, <1, 1>, <6, 6>, <7, 7>, <4, 4>, <5, 5>, <2, 2> is also random.

Now let's fix the sequence U while the sequence V is shifted to the left by the circular technique by one position.

$$U_0 = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2$$
$$V_1 = 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2 \ 3$$

This combination of pairs gives the points of the first track <3, 0>, <0, 1>, <1,6>, <6, 7>, <7, 4>, <4, 5>, <5, 2>, <2, 3> on RP. None of these points can appear on the reverse main diagonal of a grid, since sequences U_0 and V_1 are uniform. Their uniformity follows from the determined properties of the complete twisting sequences (Deon and Menyaev, 2016b). Shifting of sequence V, provided that sequence U_0 is fixed, could be continued and so the second track on a grid might be obtained.

 $U_0 = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2$ $V_2 = 1 \ 6 \ 7 \ 4 \ 5 \ 2 \ 3 \ 0$

The second track contains the points, which also cannot occur on track 1 and track 0 of the reverse diagonal and again that is because of the properties of the complete uniform twisting sequences. In this example, only 8 options to present sequence V are possible since the 9th shift repeats the initial variation $V_8 = V_0$.

So, the shift operations for sequences comply with the corresponding varieties of the twisting sequences. In the presented example, the shifts of the congruential sequence, which comply with the congruential generation from the corresponding initial values, are considered. Thus, the congruential initial generation of sequences U_0 and V_0 initiates the creation of one RP using the complete sequence of shifts of one of the original sequences V_i while another sequence U_0 is fixed. By analyzing the result of the previous program P040301, it's easy to see that among all sequences $V_{k=10,231}$ there are all 8 congruential shifts $V_{k=0}$, $V_{k=3}$, V_{k} $= 6, V_{k=9}, V_{k=12}, V_{k=15}, V_{k=18}, V_{k=21}$ in amount of $N = 2^{w}$ $= 2^3 = 8$ random numbers having length of 3 bits in the complete sequence. As a result, it turns out that tracks $\langle U_0, V_i \rangle = \langle U_0, V_K \rangle$ form the initial plane L_0 .

Similar arguments apply to plane L_1 . If we refer again to the result of the previous program *P040301*, this example shows that plane L_1 is created using the congruential twisting tracks $\langle U_k = 0, V_{k \in [1, 4, 7, 10, 13, 16, 19, 22]} \rangle$. Plane L_2 is created by using shifts of the next twister $\langle U_k = 0, V_{k \in [2, 5, 8, 11, 14, 17, 20, 23]} \rangle$. Next is plane L_3 , but to create it we need to perform the congruential circular shift of sequence U_0 by one random step to the left, which leads to obtaining the new distribution of the random values $U_{k=3}$ along the axis U. Then, the next plane L_3 , which can be obtained with the help of tracks $\langle U_{k=3}, V_{k \in [0, 3, 6, 9, 12, 15, 18, 21]} \rangle$.

A summary of all the points obtained leads us to the conclusion that number *card* (*L*) of the complete set of RPs is defined by multiplication of two things. The first is *card*(U_{CT}) = $w \cdot N = w \cdot 2^w$, which is a quantity of the different options of the congruential twisting forms of

sequence U; the second is $card(V_T) = w$, which is a quantity of the various twisters in sequence V including the congruential twister 0, i.e.,: $card(L) = card(U_{CT})$. $card(V_T) = wN \cdot w = w^2 N$.

Below is the program code P040101, in which the uniformity of all the random twisting planes L(w = 3) is checked. Matrix A is helpful for this task. Since in the uniform RPs each vertex is created once, after the full enumeration of all the RPs the quantity of generations of each vertex has to be equal to the amount of planes. In other words, the counters of cells of matrix A have to have the same values and moreover, they have to be equal to the quantity of the generated RPs. To clarify, the program below uses the random numbers with a bit length w = 3. The names P040401 and cP040401 are selected by chance.

using nsDeonYuliPlaneTwist32D;//twist-plane generator namespace P040401

{ class cP040401

//

{ static void Main(string[] args) { cDeonYuliPlaneTwist32D TP = new cDeonYuliPlaneTwist32D(); // random number bit length int w = 3; int $N = 1 \ll w$: // track length TP.SetW(w); TP.Start(); // generator starts Console.WriteLine(" $w = \{0\}$ N = $\{1\}$ ", w, N); Console.WriteLine(" $a = \{0\}$ $c = \{1\}$ $x0 = \{2\}$ ", TP.a, TP.c, TP.x0); uint[] u = new uint[N]; // point u-coords on track uint[] v = new uint[N]; // point v-coords on track uint[,] A = new uint[N, N];for (int i = 0; i < N; i++) for (int j = 0; j < N; j++) A[i, j] = 0; // point job coordinates on plane uint uu = 0: uint vv = 0; int plane = 0: // plane number int k = 0; // track number while (true) { uint nWU = TP.nWU, nRU = TP.nRU; uint nWV = TP.nRV, nRV = TP.nRV; for (int j = 0; j < N; j++) { TP.Next(ref uu, ref vv); // point on grid u[j] = uu;v[i] = vv;A[uu, vv]++; if (k % N == 0) plane++; // plane number Console.WriteLine("plane = {0}", plane); // next track k++; if (k < 569) continue; Console.Write(" $k = \{0,4\}$ ", k); Console.Write(" $nWU = \{0,3\}$ ", nWU); Console.Write(" $nRU = \{0,3\}$ ", nRU);

```
Console.Write("nWV = \{0,3\}", nWV);
     Console.Write("nRV = \{0,3\} ", nRV);
     Console.WriteLine();
     Console.Write(" U = "):
     for (int j = 0; j < N; j++)
       Console.Write("{0,4}", u[j]);
     Console.WriteLine();
     Console.Write(" V = ");
     for (int j = 0; j < N; j++)
       Console.Write("{0,4}", v[j]);
     Console.WriteLine();
     if (k \% N == 0)
     { Console.WriteLine("Matrix A");
        for (int i = 0; i < N; i++)
        { for (int j = 0; j < N; j++)
             Console.Write("{0,4}", A[i,j]);
          Console.WriteLine();
        }
     Console.ReadKey();
                               // regular result viewing
   }
 }
}
```

After executing the program *P040401*, the following listing below appears on the monitor. The skipped strings are indicated by a dashed line.

```
w = 3 N = 8
a = 5 c = 1 x0 = 3
plane = 1
k = 1 nWU = 0 nRU = 0 nWV =
                                    0 \text{ nRV} = 0
       U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
       V = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
plane = 1
k = 2 nWU = 0 nRU = 0 nWV = 1 nRV = 1
       U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
       V = 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2 \ 3
- - - - -
plane = 1
    8 nWU = 0 nRU = 0 nWV = 7 nRV = 7
\mathbf{k} =
       U = 3 \ 0 \ 1 \ 6 \ 7 \ 4 \ 5 \ 2
       V = 2 3 0 1 6 7 4 5
Matrix A
 1 \ 1 \ 1 \ 1 \ 1 \ 1
                  1
                     -1
 1 1 1 1 1 1
                     1
                   1
  1 1 1 1 1
                1
  1
    1
       1 1
             1
  1
   1 1 1 1 1
 1 1 1 1 1 1 1 1
 1 1 1 1 1 1 1 1
 1 1 1 1 1 1 1 1
- - - - -
plane = 72
```

k = 575 nWU = 2 nRU = 7 nWV = 6 nRV = 6U = 2 5 1 4 0 7 3 6 $V = 6 \ 2 \ 5 \ 1 \ 4 \ 0 \ 7 \ 3$ plane = 72k = 576 nWU = 2 nRU = 7 nWV =7 nRV = 7U = 2 5 1 4 0 7 3 6V = 2 5 1 4 0 7 3 6Matrix A 72

So, the values of the elements of matrix A confirm the analytical calculations for $card(L) = w^2N$ and for arbitrary $w \le 32$. Generator *nsDeonYuliPlaneTwist32D* creates a complete set of uniform twisting RPs, which is considered as the primary task of this article.

Conclusion

Analysis of the sources indicates that algorithms of the generators of uniform planes do not take into account the potential of the sequences having absolutely uniform distribution. Techniques of those generating algorithms do not guarantee the absolute uniformity of the complete random planes. To overcome this limitation, we proposed here the generators of the complete uniform sequences, which include unique twisting techniques described in our previous works. However, their direct application for the described task is hampered by the required properties of Descartes uniform planes. To satisfy this requirement, a new class nsDeonYuliStwist32D was constructed and now with its help it is possible to create the dynamic objects of the simplest twisters without using congruential arrays. Applying secondary indexing technique allows for getting a generator of Descartes twisting random planes, which ensures the completeness and uniqueness of all the random variables on a grid of the Descartes plane. The performed tests confirm the absolute uniform distribution of the generated random values on a plane. In addition, a variety of the initial twisting sequences allows getting a set of the twisting planes for each pair of the congruential constants. In perspective, the obtained results can be used in a large number of applied tasks, which use the spatial plane distributions.

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Author's Contributions

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Ethics

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