Sampling with Halton Points on n-Sphere

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Abstract

Motivation and Applications

Review of Low Discrepancy Sequence

Our approach

Numerical Experiments

Conclusions



Abstract



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- Sampling on n-sphere (S^n) has a wide range of applications, such as:
 - Spherical coding in MIMO wireless communication
 - ▶ Multivariate empirical mode decomposition
 - ► Filter bank design
- ▶ We propose a simple yet effective method which:
 - ▶ Utilizes low-discrepancy sequence
 - ► Contains only a few lines of Python code in our implementation!
 - ▶ Allow incremental generation.
- ▶ Numerical results show that the proposed method outperforms the randomly generated sequences and other proposed methods.



Motivation and Applications



Problem Formulation

Desirable properties of samples over S^n

- ▶ Uniform
- **▶** Deterministic
- ▶ Incremental
 - ▶ The uniformity measures are optimized with every new point.
 - Reason: in some applications, it is unknown how many points are needed to solve the problem in advance



Motivation

- ▶ The topic has been well studied for sphere in 3D, i.e. n=2
- \blacktriangleright Yet it is still unknown how to generate for n > 2.
- ▶ Potential applications (for n > 2):
 - ▶ Robotic Motion Planning (S^3 and SO(3)) (Yershova et al. 2010)
 - Spherical coding in MIMO wireless communication (Utkovski and Lindner 2006):
 - ► Cookbook for Unitary matrices
 - ightharpoonup A code word = a point in S^n
 - Multivariate empirical mode decomposition (Rehman and Mandic 2010)
 - Filter bank design (Mandic and others 2011)



Halton Sequence on S^n

- ▶ Halton sequence on S^2 has been well studied (Cui and Freeden 1997) by using cylindrical coordinates.
- \blacktriangleright Yet it is still little known for S^n where n > 2.
- ▶ Note: The generalization of cylindrical coordinates does NOT work in higher dimensions.



Review of Low Discrepancy Sequence



- ightharpoonup Generate a low discrepancy sequence over [0,1]
- ▶ Denote vdc(k, b) as a Van der Corput sequence of k points, where b is the base of a prime number.

Figure 1: Example of Van der Corput sequence



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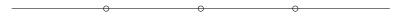


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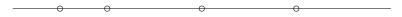


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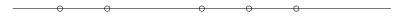


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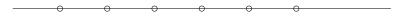


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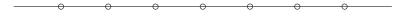


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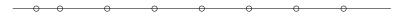


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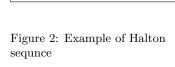
Python code

```
def vdc_basic(n, base=2):
    vdc, denom = 0.0, 1.0
    while n:
        denom *= base
        n, remainder = divmod(n, base)
        vdc += remainder / denom
    return vdc
def vdc(n, base=2):
    n - number of vectors
    base - seeds
    111
    for i in range(n):
        yield vdc basic(i, base)
```



- ► Halton sequence: using 2 Van der Corput sequences with different bases.
- Example:

$$[x,y] = [\operatorname{vdc}(k,2), \operatorname{vdc}(k,3)]$$





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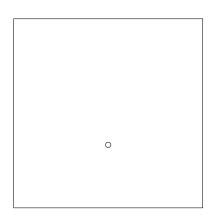


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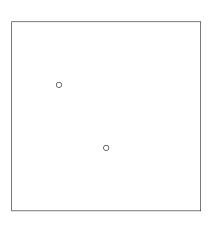


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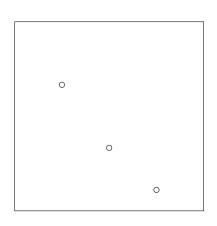


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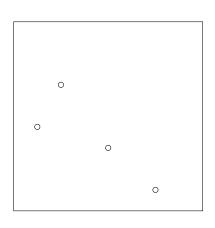


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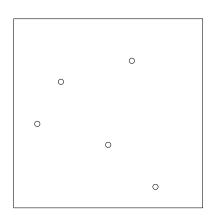


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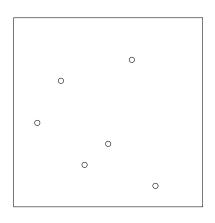


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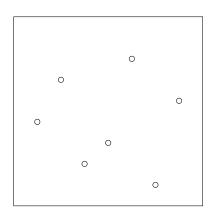


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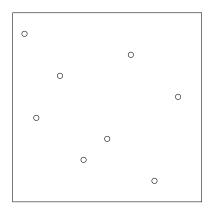


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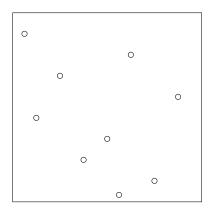


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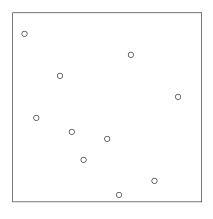


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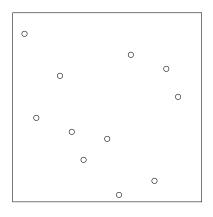


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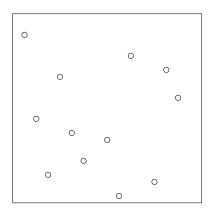


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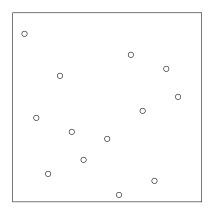


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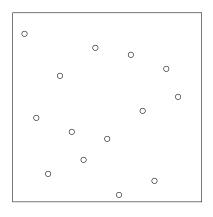


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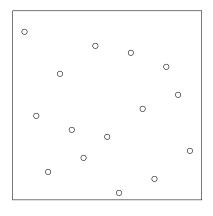


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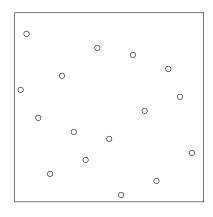


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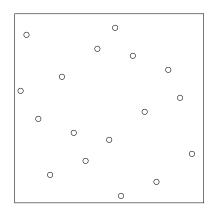


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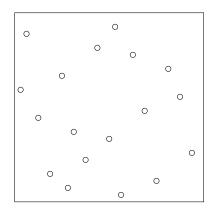


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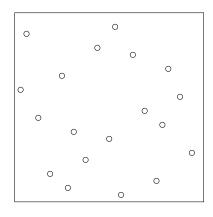


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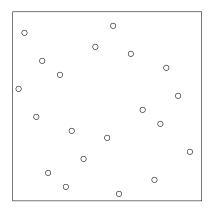


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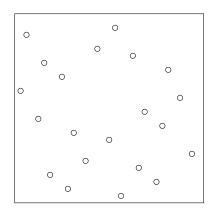


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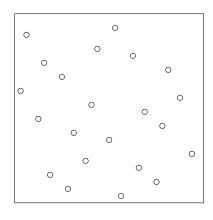


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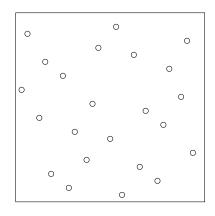


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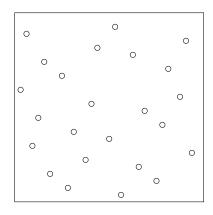


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▶ Generally we can generate Halton sequence in a unit hypercube $[0,1]^n$:

$$[x_1, x_2, \dots, x_n] = [\operatorname{vdc}(k, b_1), \operatorname{vdc}(k, b_2), \dots, \operatorname{vdc}(k, b_n)]$$

▶ A wide range of applications on Quasi-Monte Carlo Methods (QMC).



- $\theta = 2\pi \cdot \text{vdc}(k, b)$
- $[x, y] = [\cos \theta, \sin \theta]$

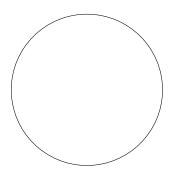


Figure 3: Sequence mapping to a unit circle



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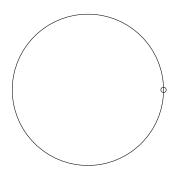


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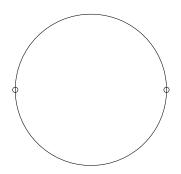


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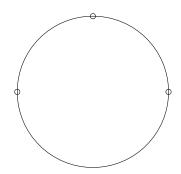


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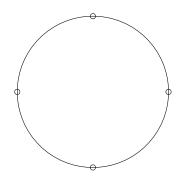


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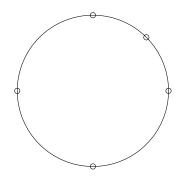


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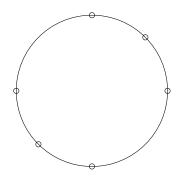


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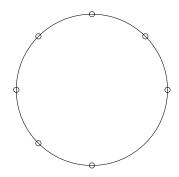


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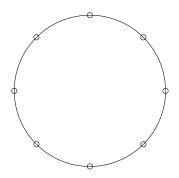


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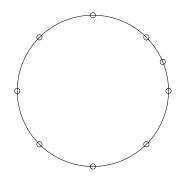


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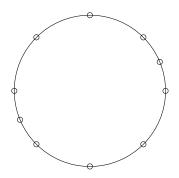


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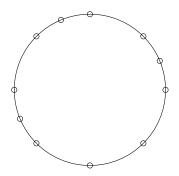


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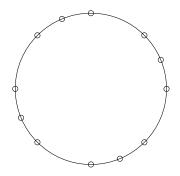


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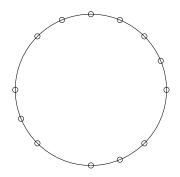


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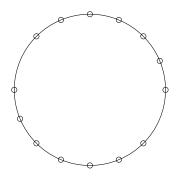


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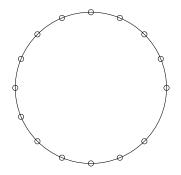


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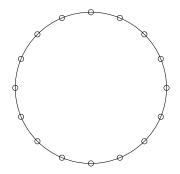


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Unit Sphere S^2

Has been applied for computer graphic applications (Wong, Luk, and Heng 1997)

- ▶ Use cylindrical mapping.
- [z, x, y] $= [\cos \theta, \sin \theta \cos \varphi, \sin \theta \sin \varphi]$ $= [z, \sqrt{1 z^2} \cos \varphi, \sqrt{1 z^2} \sin \varphi]$
- $ightharpoonup \varphi = 2\pi \cdot \operatorname{vdc}(k, b_1) \% \text{ map to } [0, 2\pi]$
- $z = 2 \cdot \text{vdc}(k, b_2) 1 \% \text{ map to } [-1, 1]$

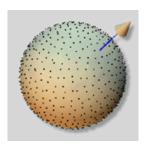


Figure 4: image



Sphere S^n and SO(3)

- ▶ Deterministic point sets
 - ▶ Optimal grid point sets for S^3 , SO(3) (Mitchell 2008; Yershova et al. 2010)
- ▶ No Halton sequences so far to the best of our knowledge.
- Note that cylindrical mapping method cannot be extended to higher dimensions.



SO(3) or S^3 Hopf Coordinates

- ► Hopf coordinates (cf. (Yershova et al. 2010))
 - $x_1 = \cos(\theta/2)\cos(\psi/2)$
 - $x_2 = \cos(\theta/2)\sin(\psi/2)$
 - $x_3 = \sin(\theta/2)\cos(\varphi + \psi/2)$
 - $x_4 = \sin(\theta/2)\sin(\varphi + \psi/2)$
- ▶ S^3 is a principal circle bundle over the S^2



Figure 5: image



Hopf Coordinates for SO(3) or S^3

Similar to the Halton sequence generation on S^2 , we perform the mapping:

- $ightharpoonup \varphi = 2\pi \cdot \operatorname{vdc}(k, b_1) \% \text{ map to } [0, 2\pi]$
- $\psi = 2\pi \cdot \text{vdc}(k, b_2)$ % map to $[0, 2\pi]$ for SO(3), or
- $\psi = 4\pi \cdot \text{vdc}(k, b_2) \%$ map to $[0, 4\pi]$ for S^3
- $z = 2 \cdot \text{vdc}(k, b_3) 1 \% \text{ map to } [-1, 1]$
- $\theta = \cos^{-1} z$



Python Code

```
def sphere3 hopf(k, b):
    vd = zip(vdc(k, b[0]), vdc(k, b[1]), vdc(k, b[2]))
    for vd0, vd1, vd2 in vd:
        phi = 2*math.pi*vd0  # map to [0, 2*math.pi]
        psy = 4*math.pi*vd1  # map to [0, 4*math.pi]
        z = 2*vd2 - 1 # map to [-1., 1.]
        theta = math.acos(z)
        cos_eta = math.cos(theta/2)
        sin eta = math.sin(theta/2)
        s = [\cos_{eta} * math.cos(psy/2),
             cos_eta * math.sin(psy/2),
             sin_eta * math.cos(phi + psy/2),
             sin_eta * math.sin(phi + psy/2)]
        vield s
```



Our approach



3-sphere

- ▶ Polar coordinates:
 - $ightharpoonup x_0 = \cos \theta_3$
 - $x_1 = \sin \theta_3 \cos \theta_2$
 - $x_2 = \sin \theta_3 \sin \theta_2 \cos \theta_1$
 - $x_3 = \sin \theta_3 \sin \theta_2 \sin \theta_1$
- ► Spherical surface element:

$$dA = \sin^2(\theta_3)\sin(\theta_2) d\theta_1 d\theta_2 d\theta_3$$



n-sphere

- ▶ Polar coordinates:
 - $x_0 = \cos \theta_n$
 - $x_1 = \sin \theta_n \cos \theta_{n-1}$
 - $x_2 = \sin \theta_n \sin \theta_{n-1} \cos \theta_{n-2}$
 - $x_3 = \sin \theta_n \sin \theta_{n-1} \sin \theta_{n-2} \cos \theta_{n-3}$
 - **...**
 - $x_{n-1} = \sin \theta_n \sin \theta_{n-1} \sin \theta_{n-2} \cdots \cos \theta_1$
 - $x_n = \sin \theta_n \sin \theta_{n-1} \sin \theta_{n-2} \cdots \sin \theta_1$
- ► Spherical surface element:

$$d^n A = \sin^{n-2}(\theta_{n-1})\sin^{n-1}(\theta_{n-2})\cdots\sin(\theta_2) d\theta_1 d\theta_2\cdots d\theta_{n-1}$$



How to Generate the Point Set

- ▶ $p_0 = [\cos \theta_1, \sin \theta_1]$ where $\theta_1 = 2\pi \cdot \text{vdc}(k, b_1)$
- ▶ Let $f_j(\theta) = \int \sin^j \theta d\theta$, where $\theta \in (0, \pi)$.
 - Note 1: $f_j(\theta)$ can be defined recursively as:

$$f_{j}(\theta) = \begin{cases} \theta & \text{if } j = 0, \\ -\cos\theta & \text{if } j = 1, \\ (1/n)(-\cos\theta\sin^{j-1}\theta + (n-1)\int\sin^{j-2}\theta d\theta) & \text{otherwise.} \end{cases}$$

- Note 2: $f_j(\theta)$ is a monotonic increasing function in $(0, \pi)$
- Map $\operatorname{vdc}(k, b_j)$ uniformly to $f_j(\theta)$: $t_j = f_j(0) + (f_j(\pi) - f_j(0))\operatorname{vdc}(k, b_j)$
- $\blacktriangleright \text{ Let } \theta_j = f_j^{-1}(t_j)$
- ▶ Define p_n recursively as: $p_n = [\cos \theta_n, \sin \theta_n \cdot p_{n-1}]$



Numerical Experiments



Testing the Correctness

- ▶ Compare the dispersion with the random point-set
 - ► Construct the convex hull for each point-set
 - Dispersion roughly measured by the difference of the maximum distance and the minimum distance between every two neighbour points:

$$\max_{a\in\mathcal{N}(b)}\{D(a,b)\}-\min_{a\in\mathcal{N}(b)}\{D(a,b)\}$$
 where $D(a,b)=\sqrt{1-a^\top b}$



Random sequences

- ightharpoonup To generate random points on S^n , spherical symmetry of the multidimensional Gaussian density function can be exploited.
- ▶ Then the normalized vector $(x_i/\|x_i\|)$ is uniformly distributed over the hypersphere S^n . (Fishman, G. F. (1996))



Convex Hull with 600 points

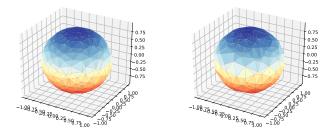


Figure 6: image

Left: our, right: random



Result for S^3

Compared with Hopf coordinate method.

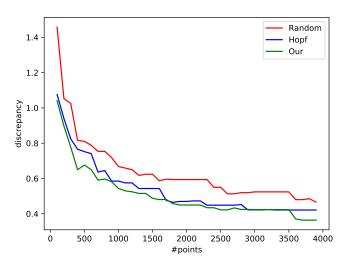


Figure 7: image



Result for S^3 (II)

Compared with cylindrical mapping method.

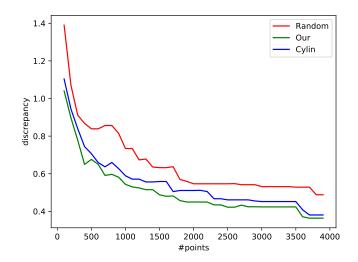


Figure 8: image



Result for S^4

Compared with cylindrical mapping method

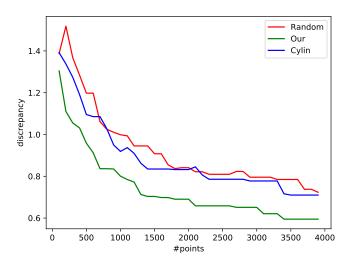


Figure 9: image



Conclusions



Conclusions

- ▶ Proposed method generates low-discrepancy point-set in nearly linear time
- ► The result outperforms the corresponding random point-set, especially when the number of points is small
- ▶ Python code is available at here



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