### 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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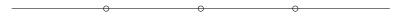


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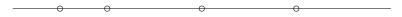


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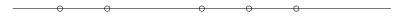


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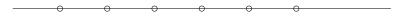


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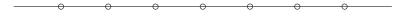


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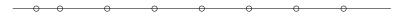


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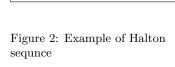
### Python code

```
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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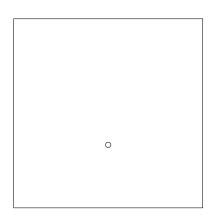


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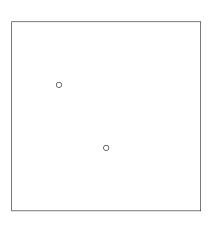


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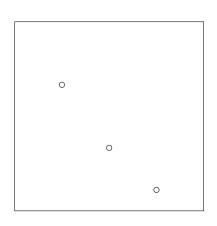


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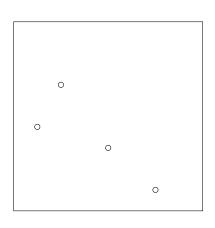


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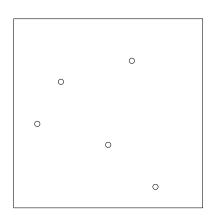


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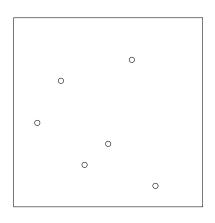


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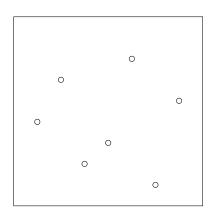


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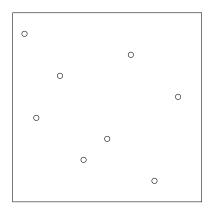


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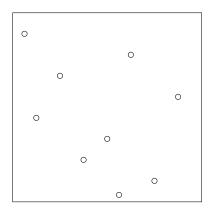


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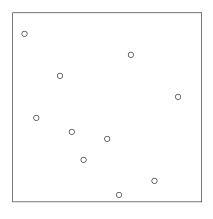


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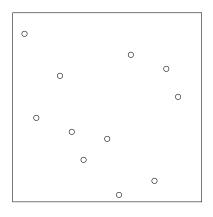


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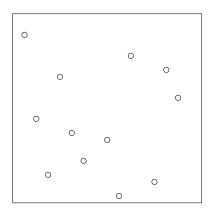


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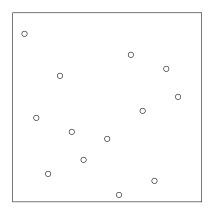


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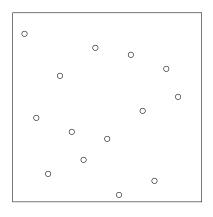


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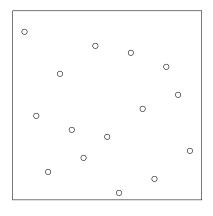


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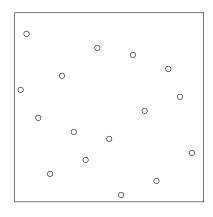


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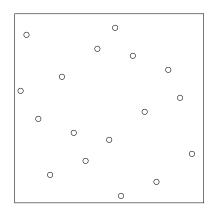


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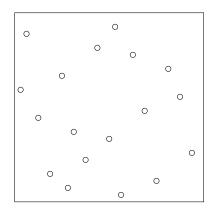


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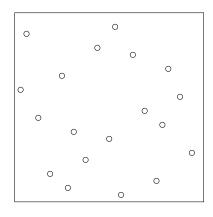


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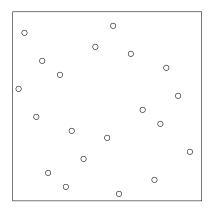


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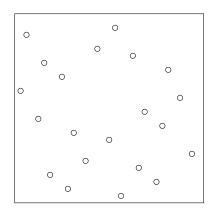


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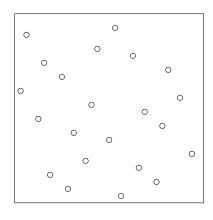


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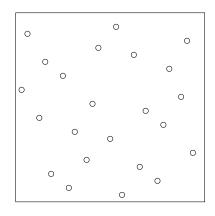


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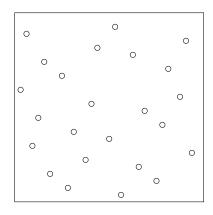


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▶ Generally we can generate Halton sequence in a unit hypercube  $[0,1]^n$ :

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

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



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

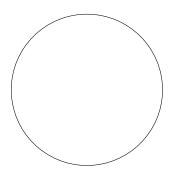


Figure 3: Sequence mapping to a unit circle



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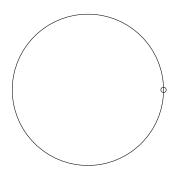


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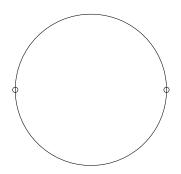


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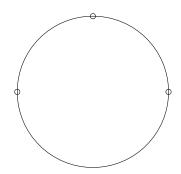


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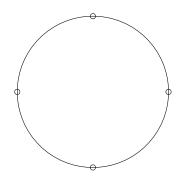


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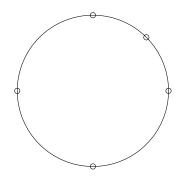


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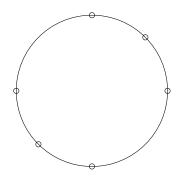


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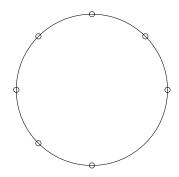


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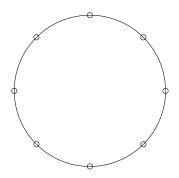


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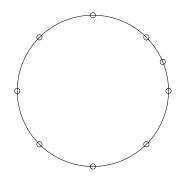


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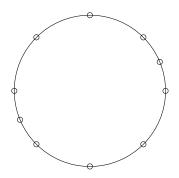


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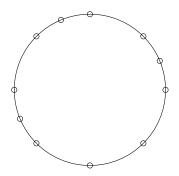


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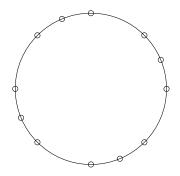


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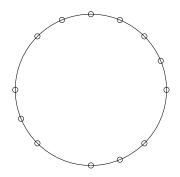


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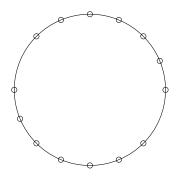


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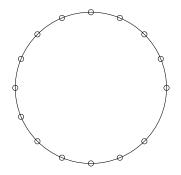


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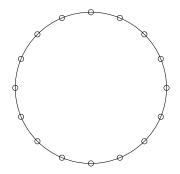


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# Unit Sphere $S^2$

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

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

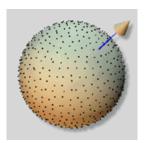


Figure 4: image



# Sphere $S^n$ and SO(3)

- ▶ Deterministic point sets
  - $\blacktriangleright$  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



# Our approach



# 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
```



### 3-sphere

- ▶ Polar coordinates:
  - $x_0 = \cos \theta_3$
  - $x_1 = \sin \theta_3 \cos \theta_2$
  - $x_2 = \sin \theta_3 \sin \theta_2 \cos \theta_1$
  - $x_3 = \sin \theta_3 \sin \theta_2 \sin \theta_1$



### n-sphere

#### ▶ Polar coordinates:

- $x_0 = \cos \theta_n$
- $x_1 = \sin \theta_n \cos \theta_{n-1}$
- $x_2 = \sin \theta_n \sin \theta_{n-1} \cos \theta_{n-2}$
- $x_3 = \sin \theta_n \sin \theta_{n-1} \sin \theta_{n-2} \cos \theta_{n-3}$
- **...**



#### 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:

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

- 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

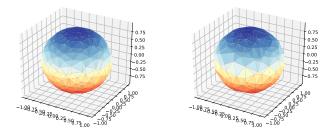
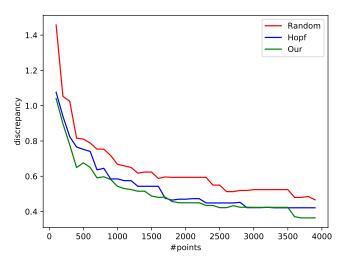


Figure 6: image

Left: our, right: random



# Result for $S^3$



 $Figure \ 7: \ image$ 



### Result for $S^4$

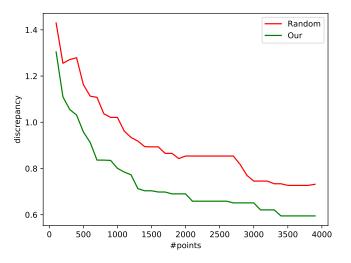


Figure 8: image



## 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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