

A Generalized 2D and 3D Hilbert Curve

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Abstract

The two and three dimensional Hilbert curves are fractal space filling curve that map the unit interval to the unit square or unit cube. Hilbert curves preserve locality, keeping distance from the unit interval in their respective higher dimensional embedding and, for this reason, find applications in scene rendering to job scheduling. Finite approximations of the Hilbert curve can be constructed in stages by stopping the recursive construction at a specified depth. One limitation of the Hilbert curve approximation construction is that the regions need to be exact powers of two, making it difficult to work in many real world scenarios. In this paper, we present the Gilbert curve, a conceptually straight forward generalization of the Hilbert curve that works on arbitrary rectangular regions. The construction provides reasonable worst case run-time guarantees for random access forward and back lookup from the unit interval to the embedded 2 or 3 dimensional space. We also provide comparisons to other Hilbert curve generalization methods and investigate overall quality metrics of the Gilbert curve construction.

1. Introduction

1.1. Overview

We present a generalized Hilbert curve for 2D and 3D curves that works on arbitrary rectangular regions. We call our realization of the generalized Hilbert curve a *Gilbert curve*.

An overview of the benefits of the Gilbert curve are that it is:

- Valid on arbitrary rectangular regions
- Equivalent to the Hilbert curve when dimensions are exact powers of 2
- Efficient at random access lookup
- Algorithmically conceptually straight forward
- Able to realize curves without notches unnecessarily and limits the resulting curve to a single notch when forced
- Similar in measures of locality to the Hilbert curve

Some drawbacks are that:

- Our implementation involves a recursive solution for random access lookup which might be too slow for applications that require better than $O(\lg N)$ runtime and memory usage
- Might not adequately capture some features of a Hilbert curve

Further, we show:

- Measures of locality are preserved (Section X)
- Trivial extensions to create generalized Moore curves (*Gore curve*, Section X)
- Other metrics to quantify qualities of the curve and their comparison to some other space filling curves

Generalizations of the Hilbert curve to non power of two rectangular cuboid regions has been explored before but are overly complicated algorithmically, create unbalanced curves and often don't generalize well to 3D. Our realization focuses on algorithmic conceptual simplicity, provides balanced resulting curves and works in both 2D and 3D.

Space filling curves are a specialization of a more general Hamiltonian path, but have a more stringent requirement

of local connectivity. The local connectivity, or locality, preserves a measure of nearness, where points from the source unit line are near in the embedded space.

The local connectivity requirements preclude things like zig-zag Hamiltonian paths, as nearby points in the embedded dimension can be far from the origin line.

The Gilbert curve algorithm works by choosing sub rectangular cuboid regions, or blocks, to recursively find a connecting path. During the course of subdivision, if the cuboid block regions stray too far from being cube like, a simpler subdivision is done to try and bring the subdivision closer to being cube like. The subdivision is done until a base case is reached and the lowest unit of the curve can be realized.

For a specified path start and end position with certain side dimensions, a Hamiltonian path is not always possible. In such a case, the Gilbert curve will introduce a single diagonal path move, called a *notch*.

1.2. Definitions

We concern ourselves with a space filling curve, ω , through a rectangular region (N_x, N_y, N_z) , $N = (N_x \cdot N_y \cdot N_z)$:

$$\begin{aligned} k &\in \{0 \dots (N-1)\}, \\ x_k, y_k, z_k &\in \mathbb{N} \\ \omega_k &= (x_k, y_k, z_k) \\ \omega &= (\omega_0, \omega_1, \dots, \omega_{N-1}) \\ k > 1 &\rightarrow |\omega_{k-1} - \omega_k| = 1 \\ \forall i, j &\in \{0 \dots (N-1)\}, i \neq j \\ &\rightarrow \omega_i \neq \omega_j \end{aligned}$$

2. Related Work

3. Algorithm

4. Experiments and Results

5. Conclusion

References

A. Appendix


A.1. Defect


Call the *defect* a function $\lambda_d : \mathbb{N}^d \mapsto \mathbb{N}$:

$$\begin{aligned} \lambda_2(w, h) &= \frac{w \cdot h}{\min(w, h)^2} \\ \lambda_3(w, h, d) &= \frac{w \cdot h \cdot d}{\min(w, h, d)^3} \end{aligned}$$

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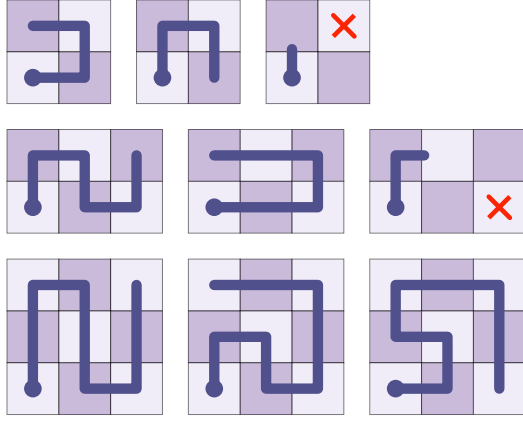


Figure 1: Illustrative examples of Hamiltonian paths height/width that are even/even, even/odd and odd/odd, respectively, when starting from the lower left hand corner

If there is a disjoint subdivision of a volume V_0 to $V_1 = (V_{0,0}, V_{0,1}, \dots, V_{0,m-1})$, $V_0 = \cup_k V_{0,k}$, define the *average defect* of the subdivided volume to be:

$$\lambda_s(V_1) = \sum_k \frac{\text{Vol}(V_{0,k})}{\text{Vol}(V_0)} \cdot \lambda(V_{0,k})$$

This weights the defect of each subdivided cuboid by its proportional volume.

The defect gives us a coarse idea of how lopsided or *eccentric* a cuboid region is. If the defect is too high, we might want to split the larger sides while keeping the smaller sides the same size.

A.2. Eccentric Split Threshold

Calculations in this section will justify what threshold value to pick of when to choose an eccentric split over a J-split. Our concern is to find a simple ratio of when each of w , h or d are considered "small enough" or "large enough", relative to the other dimensions, to split on.

An enumeration of what conditions lead to an eccentric split are as follows:

$$\begin{aligned} w &\gg h \sim d(1) \\ h &\gg w \sim d(2) \\ d &\gg w \sim h(3) \\ h &\ll w \sim d(4) \\ w &\ll h \sim d(5) \\ d &\ll w \sim h(6) \end{aligned}$$

A representation of the eccentric splits are enumerated in figure ???. The eccentric split differs from the J-split as it's only splitting in one or two dimensions, not the full three that the J-split would be doing.

For each of the six cases, we want to know what relative difference in sizes should be used to determine when an eccentric split is used and how to subdivide the cuboids so as to make progress.

Specifically, we want to find the ratio, σ or η , of when one dimension is proportionally larger or smaller, respectively, than the others and the ratio, ρ , of where to choose the split point of subdivision. For simplicity, we might want to subdivide at the half way point ($\rho = \frac{1}{2}$) but as we will

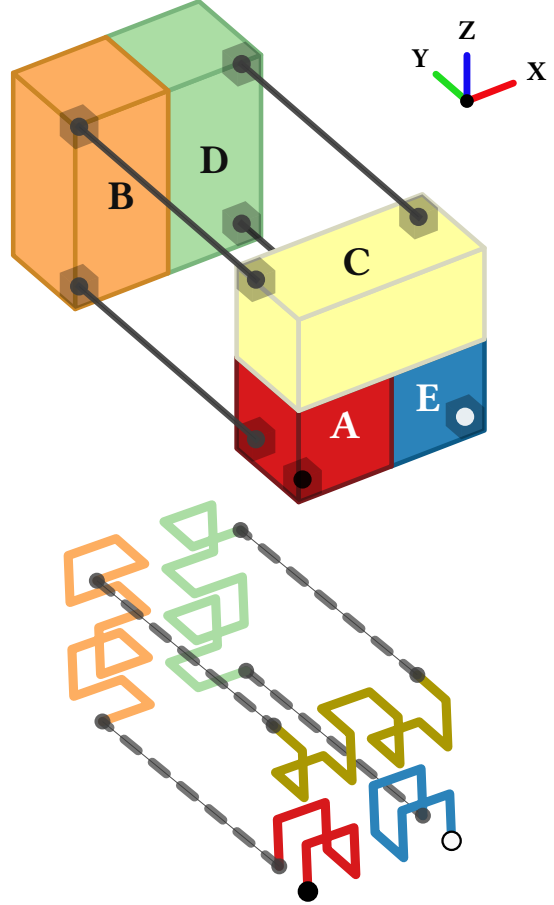


Figure 2: The J-split subdivision, representing the main subdivision of the build recursion for the 3D Gilbert curve case

see, this might give lopsided sub-cuboid regions and using a better split point is desirable.

In what follows, our goal is to choose a subdivision that will reduce the average defect. We assume that the start and end of the path lie in the w dimension with the local start point at $(0, 0, 0)$ and endpoint at $(w - 1, 0, 0)$.

Because we want to avoid adding unnecessary notches, we are forced split in the w axis. We commit to always splitting the w axis in two. If we want to subdivide into three cuboid regions, we also need to pick the split point in the other dimension in another dimension

A.3. $w \gg h \sim d$

Take $w = \sigma s$ and $h = d = s$. We commit to splitting the width dimension in half, subdividing the original volume into two equal volumes.

The defect of the original volume is $\lambda(w, h, d) = \sigma$.

If $w > 2h = 2s$, then $\min(\frac{w}{2}, h, d) = s$ and we have an average defect $\lambda_s(V_1) = \frac{\sigma}{2}$. Intuitively, this is validation that if the width is more than twice the length of the depth and height, we make progress if we split the width in half and recursively process each sub-cuboid.

If $w < 2h$ but $\min(\frac{w}{2}, h, d) = \frac{w}{2}$, the average defect $\lambda_s(V_1) = 2(\frac{1}{2})\frac{\frac{\sigma}{2} \cdot s^3}{(\frac{\sigma}{2}s)^3} = \frac{4}{\sigma^2}$. We want to reduce the average defect relative to the original defect, so:

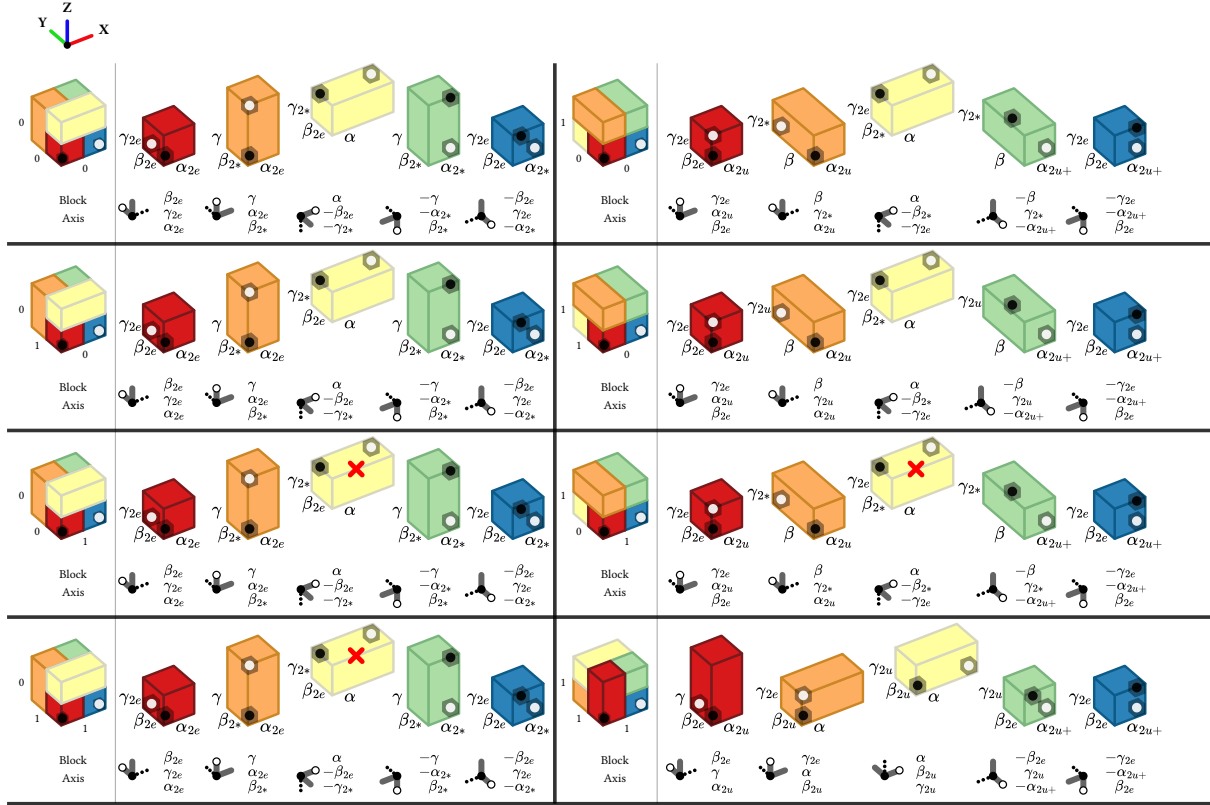


Figure 3: Bulk recursion J-split atlas for the 3D Gilbert algorithm

$$\begin{aligned}
 & \lambda_s(V_1) < \lambda(V_0) \\
 \rightarrow & \frac{4}{\sigma^2} < \sigma \\
 \rightarrow & \sigma > 4^{\frac{1}{3}} \\
 \rightarrow & \frac{5}{3} > \sigma > 4^{\frac{1}{3}} = 1.58740 \dots
 \end{aligned}$$

We've chosen the constant $\frac{5}{3}$ as a simple fraction above $4^{\frac{1}{3}}$ to know when to split w .

For this case, $\sigma = \frac{5}{3}$ and $\rho = \frac{1}{2}$. That is, we split the w axis by half when the width axis exceed $\frac{5}{3}$ of both the depth, d , and height, h , dimension.

A.4. $h \gg w \sim d$