

# Volume of a Convex Set

February 2, 2018

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- Recap: Alg of enclosing convex body  $K$  in a rectangular solid  $R$  and estimating proportion of random points from  $R$  (easy to draw) which fall in  $K$  does not work in general.
- But: the volume estimation problem for convex bodies in  $\mathbf{R}^n$  can be reduced to (nearly) uniform sampling from convex sets in  $\mathbf{R}^n$ .

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- Main Result: Given these, can estimate volume of  $K$  to relative error  $\varepsilon$  in time poly in  $n \ln(R/r)/\varepsilon$ .

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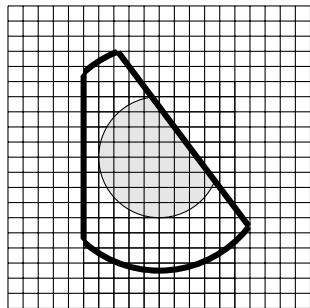
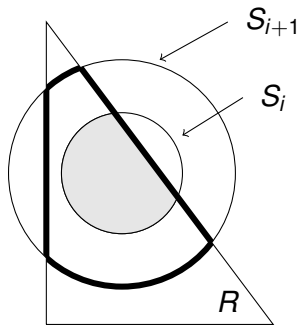
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- The ratio could be exponential in Radius of  $S_k$  / Radius of  $S_{k-1}$ .





**Figure:** By sampling the area inside the dark line and determining the fraction of points in the shaded region we compute  $\frac{\text{Vol}(S_{i+1} \cap R)}{\text{Vol}(S_i \cap R)}$ .

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- So, we have give a poly time reduction of volume estimation to random sampling from a convex set.
- Don't need exact uniform sampling - approximate (in Total Variation dist) is all we know how to do by MCMC. Is that enough? Care: Now additive error. But OK, not done explicitly here.

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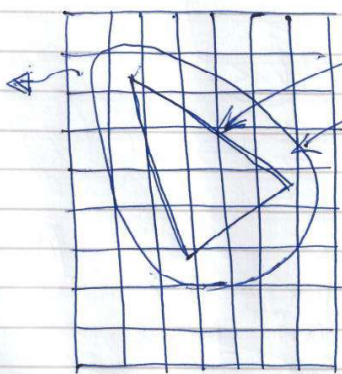
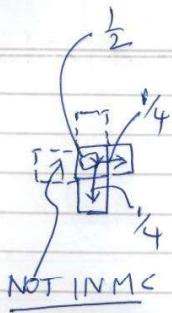
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- Uniform (symmetric MC)



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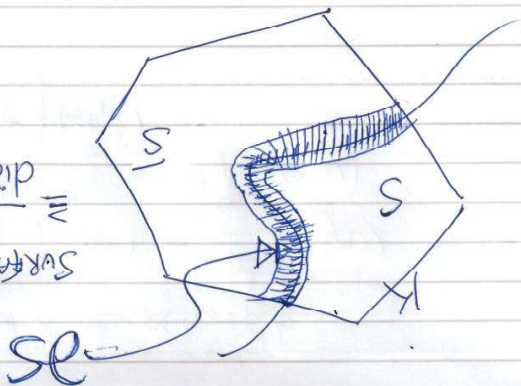
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- What is best possible  $\gamma$  ? Can't beat  $O(1/\text{diameter of } K)$ . Will prove this. [Central Result for bounding  $\Phi$ .]

$$\text{SURFACE AREA } (\partial S) \equiv \frac{1}{\text{diameter}(K)} \min(\text{Vol}(S), \text{Vol}(K))$$



# Proof of Isoperimetry - I

- Lower dimensional surface  $\partial S$  presents continuity issues. Instead, let  $S_2 = \{\mathbf{x} \in K : \text{dist}(\mathbf{x}, \partial S) \leq \varepsilon\}$  [Think of  $\varepsilon$  as small-infinitesimal.]

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- Isoperimetry Restated: For any partition of a convex into three pieces:  $S_1, S_2, S_3$  with  $\text{dist}(x, y) \geq \varepsilon \forall x \in S_1, y \in S_3$ , prove  $\text{Vol}(S_2) \geq \gamma \varepsilon \min(\text{Vol}(S_1), \text{Vol}(S_3))$ .

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- The central part of the proof of Isoperimetry is to reduce this to a 1-dimensional case and then prove (for contradiction) that for any 1-d case, these two inequalities cannot hold.

## Reduction to “Needle-like” case

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  - $P$  is nearly 1-dimensional:  $P \subseteq \varepsilon$ -neighbourhood of line segment  $a \rightarrow b$ .

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  - $P$  is nearly 1-dimensional:  $P \subseteq \varepsilon$ -neighbourhood of line segment  $a \rightarrow b$ .
  - $\int_{x=a}^b g(x)p(x), \int_{x=a}^b h(x)p(x) > 0$ , where, for  $x \in ab$ ,  $p(x)$  is the area of  $P \cap H(x)$ , with  $H(x)$  the hyperplane perpendicular to  $ab$  through  $x$ .

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- Idea of proof: Use (simple version of) Borsuk-Ulam theorem to cut the domain of integration repeatedly, preserving  $\int g, \int h > 0$



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- Choose  $V$  to be the space spanned by the two largest axes of the min vol ellipsoid containing  $K_i$ . This process will ensure that no two axes of  $K_i$  can both remain long as  $i \rightarrow \infty$ , hence,  $K_i$  become needle-like.

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$$\int_{x \in (ab) \cap S_2} p(x) \geq \min(\int_{x \in S_1} p(x), \int_{x \in S_3} p(x)) / \text{dia}(P) - \varepsilon',$$
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- Follows from the fact that the cross-sectional area  $p(x)$  is unimodal - i.e., does not decrease and then increase. So, for any  $x \in (ab) \cap S_2$ ,  $p(y) \leq p(x)$  either for all  $y > x$  or for all  $y < x$ .



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  - **Inertial Ellipsoid** Moment-of-inertia: Matrix  $M$  with  $M_{ij} = E_K(x_i x_j)$ . For  $\tau = M^{-1/2}$ ,  $E_{\tau K}(x_i x_j) = B$ .
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