

$$\lim_{n \rightarrow \infty} \Pr[|S_n| \geq t\sqrt{n}] = \sqrt{\frac{2}{\pi}} \int_t^{\infty} e^{-x^2/2} dx \leq$$

$$\leq \sqrt{\frac{2}{\pi}} \cdot \frac{1}{t} e^{-t^2/2}$$

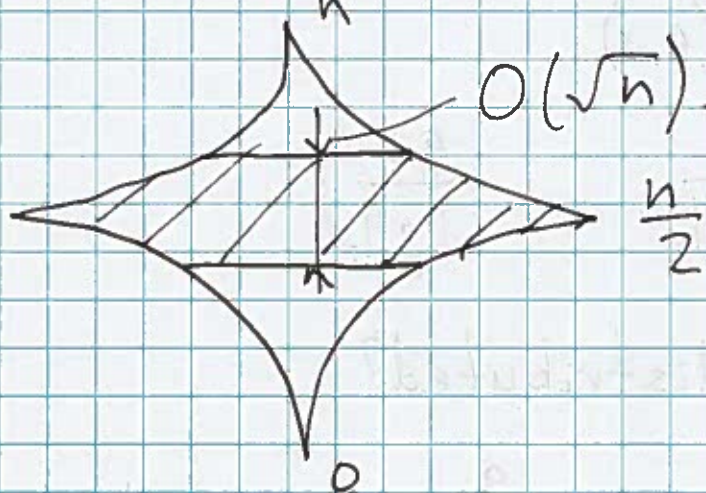
$$\geq 3\sqrt{n} \quad 0.002$$

Geometry

1) $\{0,1\}^n$

Hamming weight of a random point $= \frac{n + S_n}{2}$

99% points have Hamming weight $\frac{n}{2} \pm O(\sqrt{n})$



$$2) S^{d-1} = \{x \in \mathbb{R}^d \mid \|x\|_2 = 1\}$$

(2)

can be parametrized by $d-1$ numbers.

uniform distribution over $S^{d-1} : \mu(\cdot)$

Can be defined by:

$$1) \mu(S^{d-1}) = 1$$

$$2) \forall \text{ measurable } A \subset S^{d-1}$$

orthogonal $U \in \mathbb{R}^{d \times d}$

$$\mu(A) = \mu(UA).$$

How to sample?

$$g_1, \dots, g_d \sim \mathcal{N}(0, 1).$$

$$\text{output } \left(\frac{g_1}{\|g\|_2}, \dots, \frac{g_d}{\|g\|_2} \right)$$

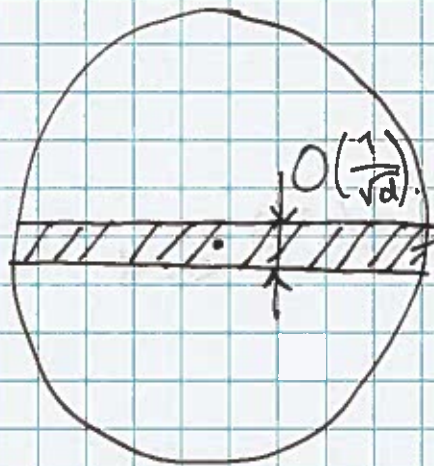
How is x_1 distributed?

$$x_1 = \frac{g_1}{\sqrt{g_1^2 + \dots + g_d^2}} = \frac{g_1}{\chi(d)} \begin{matrix} \uparrow \\ \text{dependent} \\ \downarrow \end{matrix}$$
$$\approx \frac{\mathcal{N}(0, 1)}{\sqrt{d}}, \text{ since } \chi(d) \text{ is}$$

sharply concentrated around \sqrt{d} .

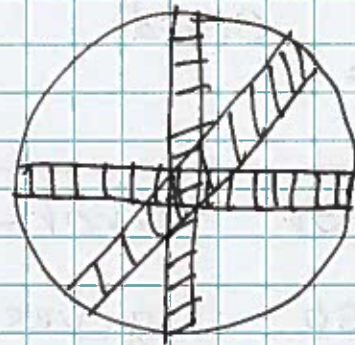
(3)

$\Pr \left[|x_1| \geq \frac{3}{\sqrt{d}} \right]$ is very small.



99%

true for any equator.



counter-intuitive!

Application

$$x_1, \dots, x_k \in S^{d-1}$$

$$\langle x_i, x_j \rangle = 0.$$

Cl $k \leq d$.

What if $|\langle x_i, x_j \rangle| \leq \epsilon$? ($\epsilon > 0$ is a small parameter).

Lm k can be $2^{\Omega(\epsilon^2 d)}$!!!

Exponentially many near-orthogonal vectors!

Pf probabilistic method

n random vectors

$$\Pr [\exists i, j : |\langle \mathbf{u}_i, \mathbf{u}_j \rangle| \geq \varepsilon] \leq$$

$$\leq n^2 \cdot \Pr [|\langle \mathbf{u}_1, \mathbf{u}_2 \rangle| \geq \varepsilon] =$$

$$= n^2 \cdot \Pr [|x_1| \geq \varepsilon] \approx n^2 e^{-\Omega(\varepsilon^2 d)} \ll 1$$

$$\text{if } n = 2^{O(\varepsilon^2 d)}$$

Useful for error-correcting codes,
but also gives JL for a regular
simplex.

Lm (JL for simplex).

n equidistant points in \mathbb{R}^{n-1} .

Can reduce dimension to $O\left(\frac{\log n}{\varepsilon^2}\right)$.