CSCI B609: "Foundations of Data Science"

Lecture 3/4: High-Dimensional Space

Slides at http://grigory.us/data-science-class.html

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Geometry of High Dimensions

- Almost all volume near the surface:
 - Take arbitrary body $A \in \mathbb{R}^d$
 - Shrink to $(1 \epsilon)A = \{(1 \epsilon)x | x \in A\}$
 - Volume change:

$$\frac{volume((1-\epsilon)A)}{volume(A)} = (1-\epsilon)^{\mathbf{d}} \le e^{-\epsilon \mathbf{d}}$$

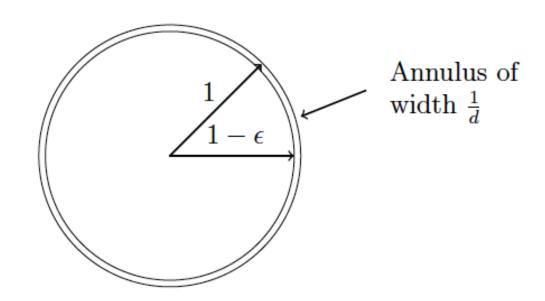
– Proof of =: partition into infinitesimal cubes

Today

- Geometry of High Dimensions (Sec 2.3 2.4)
 - Volume is near the surface
 - Volume of d-dimensional unit ball
 - Most of the volume is near equator
 - Near orthogonality of random vectors

Geometry of High Dimensions

- Let B_d = unit d-dimensional ball
- At least $1-e^{-\epsilon d}$ fraction of its volume is in the annulus of width ϵ
- $\epsilon = O\left(\frac{1}{d}\right)$: most of the volume in the annulus



- V(d) = volume of d-dimensional unit ball
- $S^d = d$ -dimensional unit sphere
- In spherical coordinates:
 - -r = radius
 - $-\Omega$ = solid angle

$$V(\boldsymbol{d}) = \int_{\Omega \in S^{\boldsymbol{d}}} \int_{\boldsymbol{r}=0}^{1} \boldsymbol{r}^{\boldsymbol{d}-1} d\boldsymbol{r} \ d\Omega = \int_{\Omega \in S^{\boldsymbol{d}}} d\Omega \int_{\boldsymbol{r}=0}^{1} \boldsymbol{r}^{\boldsymbol{d}-1} d\boldsymbol{r}$$

•
$$\int_{r=0}^{1} r^{d-1} dr = \frac{1}{d} \Rightarrow V(d) = \frac{1}{d} \int_{\Omega \in S^d} d\Omega = \frac{A(d)}{d}$$

- For
$$\mathbf{d} = 2$$
: $A(\mathbf{d}) = 2\pi \Rightarrow V(\mathbf{d}) = \pi$

- For **d** = 3:
$$A$$
(**d**) = 4π ⇒ V (**d**) = $\frac{4\pi}{3}$

•
$$A(\mathbf{d}) = \int_{\Omega \in S^{\mathbf{d}}} d\Omega = ?$$

•
$$I(\mathbf{d}) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} ... \int_{-\infty}^{+\infty} e^{-(x_1^2 + ... x_d^2)} dx_1 ... dx_d$$

In Cartesian coordinates:

$$I(\mathbf{d}) = \left[\int_{-\infty}^{+\infty} e^{-x^2} dx \right]^{\mathbf{d}} = (\sqrt{\pi})^{\mathbf{d}} = \pi^{\frac{\mathbf{d}}{2}}$$

• In spherical coordinates:

$$I(\mathbf{d}) = \int_{\Omega \in S^{\mathbf{d}}} d\Omega \int_{r=0}^{\infty} e^{-r^2} r^{\mathbf{d}-1} dr$$
$$= A(\mathbf{d}) \int_{r=0}^{\infty} e^{-r^2} r^{\mathbf{d}-1} dr$$

•
$$I(\mathbf{d}) = A(\mathbf{d}) \int_{r=0}^{\infty} e^{-r^2} r^{\mathbf{d}-1} dr$$

Let $\mathbf{r}^2 = \mathbf{t}$ (so $d\mathbf{t} = 2r d\mathbf{r} \Rightarrow d\mathbf{r} = \frac{1}{2} \mathbf{t}^{-\frac{1}{2}} d\mathbf{t}$)

•
$$\int_{r=0}^{\infty} e^{-r^{2}} r^{d-1} dr = \int_{r=0}^{\infty} e^{-t} t^{\frac{d-1}{2}} \left(\frac{1}{2} t^{-\frac{1}{2}} dt\right)$$
$$= \frac{1}{2} \int_{r=0}^{\infty} e^{-t} t^{\frac{d}{2}-1} dt = \frac{1}{2} \Gamma\left(\frac{d}{2}\right)$$

• $\Gamma(x)$ = Gamma-function (generalized factorial)

$$-\Gamma(x) = (x-1)\Gamma(x-1); \ \Gamma(1) = \Gamma(2) = 1; \ \Gamma(\frac{1}{2}) = \sqrt{\pi}$$

• We have: $I(\mathbf{d}) = (\sqrt{\pi})^{\mathbf{d}} = \frac{A(\mathbf{d})}{2} \Gamma\left(\frac{\mathbf{d}}{2}\right) \Rightarrow$

$$A(\mathbf{d}) = \frac{2 \pi^{\frac{\mathbf{d}}{2}}}{\Gamma(\frac{\mathbf{d}}{2})}; \qquad V(\mathbf{d}) = \frac{2 \pi^{\frac{\mathbf{d}}{2}}}{\mathbf{d} \Gamma(\frac{\mathbf{d}}{2})}$$

•
$$A(\mathbf{d}) = \frac{2\pi^{\frac{\mathbf{d}}{2}}}{\Gamma(\frac{\mathbf{d}}{2})}; \quad V(\mathbf{d}) = \frac{2\pi^{\frac{\mathbf{d}}{2}}}{\mathbf{d}\Gamma(\frac{\mathbf{d}}{2})}$$

• d = 2:

$$-A(2) = \frac{2\pi}{\Gamma(1)} = 2\pi; \ V(2) = \frac{\pi}{\Gamma(1)} = \pi$$

• d = 3:

$$-V(3) = \frac{2\pi^{3/2}}{3\Gamma(3/2)} = \frac{4\pi^{3/2}}{3\Gamma(1/2)} = \frac{4}{3}\pi$$
$$-A(3) = \frac{2\pi^{3/2}}{\Gamma(3/2)} = 4\pi$$

• $\Gamma\left(\frac{d}{2}\right)$ grows as a factorial of $d: \lim_{d\to\infty} V(d) = 0$

Most of the volume is near equator

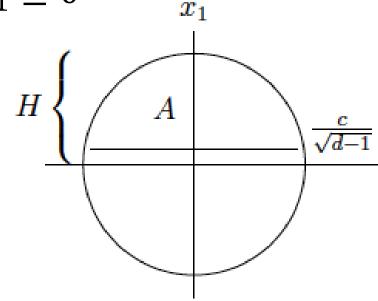
- x_1 = arbitrary coordinate
- Most of the volume has $|x_1| = O\left(\frac{1}{\sqrt{d}}\right)$
- For $c \ge 1$ and $d \ge 3$ at least $1-\frac{2}{c} e^{-\frac{c^2}{2}}$ fraction of the volume of the d-dimensional unit ball has

$$|x_1| \le \frac{c}{\sqrt{d-1}}$$

Most of the volume is near equator

- Will show: $\leq \left(\frac{2}{c} e^{-\frac{c^2}{2}}\right)$ -fraction of volume of hemisphere $x_1 \geq 0$ has $x_1 \geq \frac{c}{\sqrt{d-1}}$
- $A = \text{portion with } x_1 \ge \frac{c}{\sqrt{d-1}}$
- $H = \text{entire upper hemisphere } x_1 \ge 0$
- Will show:

$$\frac{vol(A)}{vol(H)} \le \frac{upper\ bound\ vol(A)}{lower\ bound\ vol(H)} \quad H \left\{$$



Upper bound on vol(A)

- vol(A): integrate volume of the disk of width dx_1 with face = (d-1)-dim. ball of radius $\sqrt{1-x_1^2}$
- Surface area of the disk = $(1 x_1^2)^{\frac{a-1}{2}}V(d-1)$
- $\operatorname{vol}(A) = \int_{\frac{c}{\sqrt{d-1}}}^{1} (1 x_1^2)^{\frac{d-1}{2}} V(d-1) dx_1$

• Use
$$(1-x) \le e^{-x}$$
 and $\frac{x_1\sqrt{d-1}}{c} \ge 1$:
 $vol(A) \le \int_{\sqrt{d-1}}^{\infty} \frac{x_1\sqrt{d-1}}{c} e^{-\frac{d-1}{2}x_1^2} V(d-1) dx_1 = \cdots$

$$= V(d-1)\frac{\sqrt{d-1}}{c} \times \frac{1}{d-1}e^{-\frac{c^2}{2}} = V(d-1)\frac{e^{-\frac{c^2}{2}}}{c\sqrt{d-1}}$$

Lower bound on vol(H)

- vol(H) = volume of hemisphere with $x_1 \le \frac{c}{\sqrt{d-1}}$
- $vol(H) \ge volume of hemisphere with <math>x_1 \le \frac{1}{\sqrt{d-1}}$
- $vol(H) \ge volume of cylinder with:$
 - Height: $h = \frac{1}{\sqrt{d-1}}$
 - Radius: $R = \sqrt{1 \frac{1}{d-1}}$
- Volume of cylinder = $h \times V(d-1)R^{d-1} =$

$$\frac{1}{\sqrt{d-1}}V(d-1)\left(1-\frac{1}{d-1}\right)^{\frac{d-1}{2}} \ge \frac{V(d-1)}{2\sqrt{d-1}}$$

• Last inequality since $(1-x)^a \ge 1 - ax$ (for $a \ge 1$)

Putting things together

$$\frac{vol(A)}{vol(H)} \le \frac{upper\ bound\ vol(A)}{lower\ bound\ vol(H)}$$

$$\le \frac{V(d-1)\frac{e^{-\frac{c^2}{2}}}{c\sqrt{d-1}}}{\frac{V(d-1)}{2\sqrt{d-1}}} = \frac{2e^{-\frac{c^2}{2}}}{c}$$

• **Q:** Why didn't we use $vol(H) = \frac{1}{2}V(d)$?

Today:

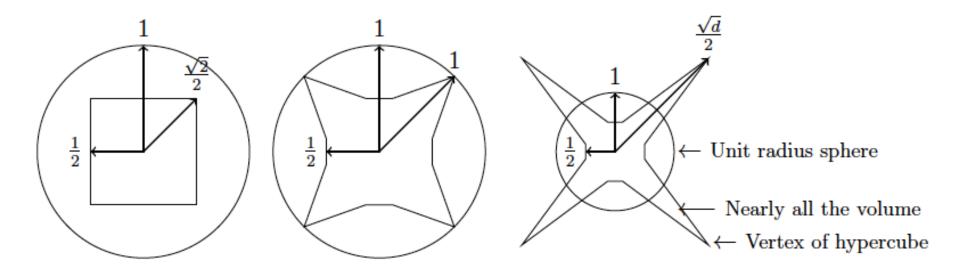
 \approx Sec 2.4.2 – 2.7

- Near orthogonality of random vectors
- Sampling Uniform Distribution over B_d
- Gaussian Annulus Theorem (concentration)
- Nearest neighbor search & random projections

Near orthogonality

- Consider drawing n points $x_1, ..., x_n$ at random from the unit d-dimensional ball
- Thm: With probability 1 O(1/n):
 - $\text{ For all } i: \left| |x_i| \right|_2 \ge 1 \frac{2 \ln n}{d}$
 - For all $i \neq j$: $\left| \langle x_i, x_j \rangle \right| \leq \frac{\sqrt{6 \ln n}}{\sqrt{d-1}}$
- $\Pr\left[\left||x_i|\right|_2 < 1 \frac{2\ln n}{d}\right] \le e^{-\frac{2\ln n}{d}d} = \frac{1}{n^2}$
- $\Pr\left[\left|\langle x_i, x_j \rangle\right| > \frac{\sqrt{6 \ln n}}{\sqrt{d-1}}\right] \le O\left(e^{-\frac{6 \ln n}{2}}\right) = O(n^{-3})$
- + Union bound (over n vectors and $O(n^2)$ pairs)

Sphere vs. cube in 2, 4, d dimensions



Sampling Uniform Distribution over B_d

- How to sample uniformly from a unit ball?
- Sample uniformly from a unit cube
 - Output the sample if inside B_d
 - Repeat if outside B_d
- Number of repetitions to output a sample?

Normal Distribution

- Normal distribution N(0,1)
 - Range: $(-\infty, +\infty)$
 - Density: $\mu(x) = (2\pi)^{-\frac{1}{2}} e^{-\frac{x^2}{2}}$
 - Mean = 0, Variance = 1
- Basic facts:
 - If X and Y are independent r.v. with normal distribution then X + Y has normal distribution
 - $-Var[cX] = c^2 Var[X]$
 - If X, Y are independent, then:

$$Var[X + Y] = Var[X] + Var[Y]$$

Sampling Uniform Distribution over B_d

- Sample $x_1, x_2 \dots, x_d$ i.i.d with $x_i \sim N(0,1)$
- $\Pr[x_i = z] = (2\pi)^{-\frac{1}{2}} e^{-\frac{z^2}{2}}$
- $\Pr[\mathbf{x} = (z_1, \dots, z_d)] = (2\pi)^{-\frac{d}{2}} e^{-\frac{z_1^2 + z_2^2 + \dots + z_d^2}{2}}$
- $\frac{x}{||x||_2} \sim U(S_d)$, how to make it $U(B_d)$?
- Scale by $\rho \in [0,1]$: $U(B_d) = \frac{\rho x}{||x||_2}$ for $\rho(r) = dr^{d-1}$

$$V(\mathbf{d}) = \int_0^1 A(\mathbf{d}) r^{\mathbf{d}-1} dr \Rightarrow 1 = \int_0^1 \frac{A(\mathbf{d})}{V(\mathbf{d})} r^{\mathbf{d}-1} dr$$

Gaussian Annulus Theorem

• Gaussian in d dimensions $(N_d(0^d, 1))$:

$$\Pr[\mathbf{x} = (z_1, \dots, z_d)] = (2\pi)^{-\frac{d}{2}} e^{-\frac{z_1^2 + z_2^2 + \dots + z_d^2}{2}}$$

Nearly all mass in annulus of radius \sqrt{d} and width O(1):

- **Thm.** For any $\beta \le \sqrt{d}$ all but $3e^{-c\beta^2}$ probability mass satisfies $\sqrt{d} \beta \le \big||x|\big|_2 \le \sqrt{d} + \beta$ for constant c
- **Proof:** Let $y = (y_1, ..., y_d) \sim N_d(0^d, 1)$ and $r = ||y||_2$

$$-|r-\sqrt{d}| \ge \beta \Leftrightarrow |r^2-d| \ge \beta(r+\sqrt{d}) \ge \beta\sqrt{d}$$

- Will bound $\Pr[|r^2 - d| \ge \beta \sqrt{d}]$

Gaussians in High Dimension

- Will bound $\Pr[|r^2 d| \ge \beta \sqrt{d}]$
- $r^2 \mathbf{d} = (y_1^2 1) + \dots + (y_d^2 1)$
- Let $x_i = y_i^2 1$, bound $\Pr[\left|\sum_{i=1}^d x_i\right| \ge \beta \sqrt{d}]$
- $\mathbb{E}[x_i] = \mathbb{E}[y_i^2] 1 = 0$
- Fix an integer s > 1
 - For $|y_i| \le 1$ we have $|x_i|^s \le 1$
 - For $|y_i| \ge 1$ we have $|x_i|^s \le |y_i|^{2s}$
- $|\mathbb{E}[x_i^s]| \le \mathbb{E}[|x_i|^s] \le \mathbb{E}[1 + y_i^{2s}] = 1 + \mathbb{E}[y_i^{2s}]$
- $1 + \mathbb{E}[y_i^{2s}] = 1 + \sqrt{2/\pi} \int_0^\infty y^{2s} e^{-\frac{y^2}{2}} dy \le 2^s s!$

Gaussians in High Dimension

Let $z = \sum_{i=1}^n z_i$ where z_i are i.i.d r.vs: $\mathbb{E}[z_i] = 0$ and $\text{Var}[z_i] \leq \sigma^2$

Thm 12.5. If $a \in [0, \sqrt{2}n\sigma^2]$ and $a^2/(4n\sigma^2) \le s \le n\sigma^2/2$, s is an even integer and $|\mathbb{E}[z_i^r]| \le \sigma^2 r!$ for all r = 3,4,...,s

$$\Pr\left[\left|\sum_{i=1}^{n} z_i\right| \ge a\right] \le 3e^{-\frac{a^2}{12n\sigma^2}}$$

- Take $a = \beta \sqrt{d}$, n = d, scale $x_i \to w_i = \frac{x_i}{2}$
- $|\mathbb{E}[x_i^s]| \le 2^s s! \Rightarrow |\mathbb{E}[w_i^s]| \le s!$

•
$$\mathbb{E}[x_i] = 0 \Rightarrow Var[w_i] = \frac{1}{4}Var[x_i] = \frac{1}{4}\mathbb{E}[x_i^2] \le \frac{2^2 2!}{4} = 2 = \sigma^2$$

$$\Pr\left[\frac{1}{2} \left| \sum_{i=1}^{n} x_i \right| \ge \beta \sqrt{d} \right] \le 3e^{-c\beta^2}$$