

Lecture 9: 25 September 2023

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9.1 Performance of DRIVE

We have seen that

$$\mathbf{R}\mathbf{x} \sim \text{Unif} \left(\mathbb{S} \left(\mathbf{0}, \|\mathbf{x}\|_2^2 \right) \right). \quad (9.1)$$

Consider

$$\mathbf{U} \triangleq \frac{\mathbf{Z}}{\|\mathbf{Z}\|_2} \quad (9.2)$$

where $\mathbf{Z} \sim \mathcal{N}(\mathbf{0}, 1)$. Then,

$$\mathbf{U} \sim \text{Unif}(\mathbb{S}(\mathbf{0}, 1)). \quad (9.3)$$

Notice that

$$f_{\mathbf{U}|\|\mathbf{Z}\|_2=r}(\mathbf{u}) = f_{\mathbf{U}}(u) = \begin{cases} \frac{1}{\text{Ar}(\mathbb{S}(\mathbf{0}, 1))} & \|u\| = 1 \\ 0 & \text{else} \end{cases} \quad (9.4)$$

so that \mathbf{U} and $\|\mathbf{Z}\|_2$ are statistically independent. Thus, from (9.2),

$$\mathbf{U} \|\mathbf{Z}\|_2 = \mathbf{Z} \quad (9.5)$$

$$\|\mathbf{U}\|_1 \|\mathbf{Z}\|_2 = \|\mathbf{Z}\|_1 \quad (9.6)$$

$$\|\mathbf{U}\|_1 = \frac{\|\mathbf{Z}\|_2}{\|\mathbf{Z}\|_1}. \quad (9.7)$$

Clearly, the 2-norm of \mathbf{Z} and 1-norm of \mathbf{U} are statistically independent.

Note that

$$\mathbb{E} \left[\|\mathbf{Z}\|_1^2 \right] = \mathbb{E} \left[\left(\sum_{i=1}^d |Z_i| \right)^2 \right] \quad (9.8)$$

$$= \mathbb{E} \left[\sum_{i=1}^d \sum_{j=1}^d |Z_i| |Z_j| \right] \quad (9.9)$$

$$= \mathbb{E} \left[\sum_{i=1}^d |Z_i|^2 + \sum_{i=1}^d \sum_{j=1, j \neq i}^d |Z_i| |Z_j| \right] \quad (9.10)$$

$$= d + \sum_{i=1}^d \sum_{j=1, j \neq i}^d \mathbb{E} [|Z_i| |Z_j|] \quad (9.11)$$

$$= d + d(d-1) \frac{2}{\pi} \quad (9.12)$$

since

$$\mathbb{E} [|Z_i|] = \int_{-\infty}^{\infty} |z_i| \frac{e^{-\frac{z_i^2}{2}}}{\sqrt{2\pi}} dz_i \quad (9.13)$$

$$= 2 \int_0^{\infty} z_i \frac{e^{-\frac{z_i^2}{2}}}{\sqrt{2\pi}} dz_i \quad (9.14)$$

$$= \sqrt{\frac{2}{\pi}} \int_0^{\infty} e^{-y} dy = \sqrt{\frac{2}{\pi}} \quad (9.15)$$

where we make the change of variables $y \triangleq \frac{z_i^2}{2}$. Using (9.12),

$$\mathbb{E} [\|\mathbf{U}\|_1^2] = \frac{\mathbb{E} [\|\mathbf{Z}\|_1^2]}{\|\mathbf{Z}\|_2^2} \quad (9.16)$$

$$= \frac{d + d(d-1) \frac{2}{\pi}}{d} = 1 + (d-1) \frac{2}{\pi}. \quad (9.17)$$

Using (9.17), for any norm $\|\mathbf{x}\|_2^2$ and taking $d \rightarrow \infty$, we obtain

$$\text{MSE} = \left(1 - \frac{2}{\pi}\right) \|\mathbf{x}\|_2^2. \quad (9.18)$$

9.2 Generating a Uniform Rotation Matrix

We can generate

$$\mathbf{A}_{d \times d} \sim \mathcal{N}(0, 1) \quad (9.19)$$

and perform Gram-Schmidt orthogonalization or take a QR -decomposition to obtain the orthonormal matrix Q .

Lemma 9.1. *If \mathbf{A} is a randomly generated $d \times d$ matrix with all entries drawn independently from the standard normal distribution, then $\mathbf{A} = \mathbf{Q}\mathbf{R}$ where \mathbf{Q} is a uniform rotation matrix.*

9.3 Structured Random Rotation Matrices

It is costly to share $\mathcal{O}(d^2)$ bits. We present an alternate choice of \mathbf{R} that does incur a higher MSE but shares less randomness. We define

$$\mathbf{R} \triangleq \frac{1}{\sqrt{d}} \mathbf{H}_l \mathbf{D} \quad (9.20)$$

where we assume that $d = 2^l$ for some nonnegative integer l and \mathbf{H}_l is the d -dimensional *Walsh-Hadamard* matrix, which is recursively defined as

$$\mathbf{H}_1 = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \quad (9.21)$$

$$\mathbf{H}_l = \begin{pmatrix} \mathbf{H}_{l-1} & \mathbf{H}_{l-1} \\ \mathbf{H}_{l-1} & -\mathbf{H}_{l-1} \end{pmatrix} \quad (9.22)$$

and \mathbf{D} is a diagonal matrix with iid Rademacher $\{1, -1\}$ entries.

Using this choice of \mathbf{R} , the overall complexity reduces to $\mathcal{O}(d \log d)$ and the MSE is still $\Theta(\|\mathbf{x}\|_2^2)$.