

## Chapter 0 Appetizer: using probability to cover geometric sets

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May 21, 2018

### Exercise 0.1.3.

#### Solution Part 1

We will first show that for independent random variables, variance of the sum is equal to sum of the variances.

$$\text{var} \left( \sum_{i=1}^n X_i \right) = \mathbb{E} \left[ \sum_{i=1}^n X_i \right]^2 - \left[ \mathbb{E} \sum_{i=1}^n X_i \right]^2 \quad (1)$$

Now consider each of the two terms on the RHS of this equation.

$$\mathbb{E} \left[ \sum_{i=1}^n X_i \right]^2 = \mathbb{E} \left( \sum_{i=1}^n \sum_{j=1}^n X_i X_j \right) = \sum_{i=1}^n \sum_{j=1}^n (\mathbb{E} X_i X_j)$$

Similarly,

$$\left[ \mathbb{E} \sum_{i=1}^n X_i \right]^2 = \left[ \sum_{i=1}^n \mathbb{E} X_i \right]^2 = \sum_{i=1}^n \sum_{j=1}^n \mathbb{E} X_i \mathbb{E} X_j$$

Using these in 1, we get

$$\text{var} \left( \sum_{i=1}^n X_i \right) = \sum_{i=1}^n \sum_{j=1}^n (\mathbb{E} X_i X_j - \mathbb{E} X_i \mathbb{E} X_j) = \sum_{i=1}^n \sum_{j=1}^n \text{cov}(X_i, X_j)$$

If the variables are independent, we can write

$$\text{var} \left( \sum_{i=1}^n X_i \right) = \sum_{i=1}^n \sum_{j=1}^n \text{cov}(X_i, X_j) = \sum_{i=1}^n \text{cov}(X_i, X_i) = \sum_{i=1}^n \text{var}(X_i)$$

Now consider the zero-mean independent random variables given. From the above proof, we

have

$$\begin{aligned}
\text{var} \left( \sum_{j=1}^k Z_j \right) &= \sum_{j=1}^k \text{var} (Z_j) \\
\mathbb{E} \left\| \sum_{j=1}^k Z_j \right\|_2^2 - \left\| \mathbb{E} \sum_{j=1}^k Z_j \right\|_2^2 &= \sum_{j=1}^k \mathbb{E} \|Z_j\|_2^2 - \sum_{j=1}^k \|\mathbb{E} Z_j\|_2^2 && \text{(From the definition of variance)} \\
\mathbb{E} \left\| \sum_{j=1}^k Z_j \right\|_2^2 - \left\| \sum_{j=1}^k \mathbb{E} Z_j \right\|_2^2 &= \sum_{j=1}^k \mathbb{E} \|Z_j\|_2^2 - \sum_{j=1}^k \|\mathbb{E} Z_j\|_2^2 && \text{(Expectation of a sum equals sum of the expectations)} \\
\mathbb{E} \left\| \sum_{j=1}^k Z_j \right\|_2^2 &= \sum_{j=1}^k \mathbb{E} \|Z_j\|_2^2 && (\mathbb{E} Z_j = 0)
\end{aligned}$$

This is the required result in Part 1.

## Part 2

This is derived easily from the linearity of expectations.

$$\begin{aligned}
\mathbb{E} \|Z - \mathbb{E} Z\|_2^2 &= \mathbb{E} \left( \|Z\|_2^2 - 2Z \mathbb{E} Z + \|\mathbb{E} Z\|_2^2 \right) \\
&= \mathbb{E} \|Z\|_2^2 - 2\mathbb{E} Z \mathbb{E} Z + \|\mathbb{E} Z\|_2^2 && \text{(Expectation of an expectation is the same)} \\
&= \mathbb{E} \|Z\|_2^2 - 2\|\mathbb{E} Z\|_2^2 + \|\mathbb{E} Z\|_2^2 && \text{(Product of same expected values)} \\
&= \mathbb{E} \|Z\|_2^2 - \|\mathbb{E} Z\|_2^2
\end{aligned}$$

which is the required result.

## Exercise 0.1.5.

**Solution** We will first show the lower bound. We have

$$\begin{aligned}
m &\leq n \\
\Rightarrow mk &\leq nk && \text{(for some } k > 0) \\
\Rightarrow mn - mk &\geq mn - nk \\
\Rightarrow \frac{n-k}{n} &\geq \frac{m-k}{m} && \text{(Dividing both sides by } mn) \\
\Rightarrow \frac{n}{m} &\leq \frac{n-k}{m-k} \\
\Rightarrow \left( \frac{n}{m} \right)^m &\leq \frac{n-1}{m-1} \cdots \frac{n-m+1}{1} = \binom{n}{m}
\end{aligned}$$

To show the upper bound, observe that

$$\binom{n}{m} = \frac{n!}{m!(n-m)!} = \frac{n(n-1)\dots(n-m+1)}{m!} \leq \frac{n^m}{m!} \quad (2)$$

Consider the Taylor expansion of  $e^m$ :

$$e^m = \sum_{k=0}^{\infty} \frac{m^k}{k!}$$

If we take only the  $m^{\text{th}}$  term on the RHS, we get

$$e^m > \frac{m^m}{m!} \Rightarrow \frac{1}{m!} < \left(\frac{e}{m}\right)^m. \quad (3)$$

Using 3 in 2, we obtain the upper bound, i.e.,

$$\binom{n}{m} < \left(\frac{en}{m}\right)^m.$$

**Exercise 0.1.6.** (Improved covering)

**Solution** This proof follows the proof of Corollary 0.1.4 very closely, and differs only at the final step.

Given that  $P$  is a polytrope with  $N$  vertices. Let centers of the required balls be defined as

$$\mathcal{N} := \left( \frac{1}{k} \sum_{j=1}^k x_j : x_j \text{ are the vertices of } P \right),$$

where  $k = \frac{1}{\epsilon^2}$ .

For any point  $x \in P$ ,  $x$  is within distance  $\frac{1}{\sqrt{k}} \leq \epsilon$  from some point in  $\mathcal{N}$ . Now we just have to show the required bound on  $|\mathcal{N}|$ . Essentially, this quantity is equal to the number of ways of choosing  $k$  elements with replacement from a set of  $N$  elements.

$$\begin{aligned} |\mathcal{N}| &= \binom{N+k-1}{k} && \leq \left( \frac{e(N+k-1)}{k} \right)^k && \text{(Result from previous problem)} \\ &= [e(1-\epsilon^2) + e\epsilon^2 N]^{\frac{1}{\epsilon^2}} && && \text{(Substituting value of k)} \\ &< [e + e\epsilon^2 N]^{\frac{1}{\epsilon^2}}, \end{aligned}$$

which is the desired result.