

1 A few notes on Lecture 14

1.1 The distance bound

Recall that we want to bound

$$|Z_i - Z_{i-1}| = |E[f(X_1, \dots, X_{i-1}, X_i, X_{i+1}, \dots, X_n) - f(X_1, \dots, X_{i-1}, \widehat{X}_i, X_{i+1}, \dots, X_n) \mid X_1, X_2, \dots, X_i]| \quad (1)$$

Note that we do not want to just use the $2\sqrt{2}$ -Lipschitz property, since that will be too weak, and will only give us

$$\Pr[|f - Ef| \leq \lambda] \leq \exp\{-\lambda^2/O(n)\}.$$

We want something much better!

Claim 1.1

$$\begin{aligned} & f(X_1, \dots, X_{i-1}, X_i, X_{i+1}, \dots, X_n) - f(X_1, \dots, X_{i-1}, \widehat{X}_i, X_{i+1}, \dots, X_n) \\ & \leq 2\left(\min_{j \neq i} d(X_i, X_j) + \min_{j \neq i} d(\widehat{X}_i, X_j)\right). \end{aligned}$$

Proof: Let $A = X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n$ be all the points except X_i and \widehat{X}_i , and let $T(A)$ be the optimal TSP tour on A . Note that $f(A) = \text{length}(T(A))$. For any point x and set S , define $d(x, S) = \min_{y \in S} d(x, y)$.

Note that if take $T(A)$, and to it we add two edges from X_i to its closest point in A , and from \widehat{X}_i to its closest point in A , then we have an Eulerian graph on the $n + 1$ points $A \cup \{X_i, \widehat{X}_i\}$ of total length at most

$$f(A) + 2(d(X_i, A) + d(\widehat{X}_i, A)). \quad (2)$$

Using the triangle inequality to shortcut repeated vertices gives us TSP tour of length at most (2), and hence the length of the optimal tour on $A \cup \{X_i, \widehat{X}_i\}$ has length

$$f(A \cup \{X_i, \widehat{X}_i\}) \leq f(A) + 2(d(X_i, A) + d(\widehat{X}_i, A)). \quad (3)$$

Finally, using the fact that

$$\begin{aligned} f(A) & \leq f(A \cup \{X_i\}) \leq f(A \cup \{X_i, \widehat{X}_i\}) \\ f(A) & \leq f(A \cup \{\widehat{X}_i\}) \leq f(A \cup \{X_i, \widehat{X}_i\}) \end{aligned}$$

implies that

$$\begin{aligned} |f(A \cup \{X_i\}) - f(A \cup \{\widehat{X}_i\})| & \leq f(A \cup \{X_i, \widehat{X}_i\}) - f(A) \\ & \leq 2(d(X_i, A) + d(\widehat{X}_i, A)), \end{aligned} \quad (4)$$

the last inequality using (3). This is just a rephrasing of the claim that we want to prove. ■

Corollary 1.2 Define $B = \{X_j \mid j > i\}$. Then

$$f(X_1, \dots, X_{i-1}, X_i, X_{i+1}, \dots, X_n) - f(X_1, \dots, X_{i-1}, \widehat{X}_i, X_{i+1}, \dots, X_n) \leq 2(d(X_i, B) + d(\widehat{X}_i, B)). \quad (5)$$

Proof: The quantity on the right of (5) is larger than the quantity on the right of (4), since $B \subseteq A$. ■

Lemma 1.3

$$|Z_i - Z_{i-1}| \leq 2(E[d(X_i, B) | X_i] + E[d(\widehat{X}_i, B)]). \quad (6)$$

Proof: Plug in the result of Claim 1.1 into (1), and note that $d(X_i, B)$ is independent of X_1, X_2, \dots, X_{i-1} , whereas $d(\widehat{X}_i, B)$ is independent of all X_1, \dots, X_i . Simplifying gives us the lemma. ■

1.2 The Rest of the Argument

Suppose we throw down $n - i$ points randomly in \mathcal{U} , and define the random variable $Q_i(x)$ to be the distance of x to the closest point amongst these $n - i$. Let R be the set of random points, and hence $Q_i(x) = d(x, R)$. We proved that

Claim 1.4 For any $x \in \mathcal{U}$,

$$E[Q_i(x)] \leq \frac{O(1)}{\sqrt{n-i}}. \quad (7)$$

Proof: This was the geometric proof, and I am going to omit it. ■

Hence we can upper bound both $E[d(X_i, B) | X_i]$ and $E[d(\widehat{X}_i, B)]$ by $\frac{O(1)}{\sqrt{n-i}}$. Finally, using (6), we get

$$|Z_i - Z_{i-1}| \leq 2\left(\frac{O(1)}{\sqrt{n-i}} + \frac{O(1)}{\sqrt{n-i}}\right) \quad (8)$$

This implies that we can set $c_i = \frac{O(1)}{\sqrt{n-i}}$ in Azuma's inequality, which is much better than the bound that we get just plugging in the $2\sqrt{2}$ -Lipschitz-ness of f . Now $\sum_i c_i^2 = O(\log n)$, and hence we get

$$\Pr[|f - Ef| \leq \lambda] \leq \exp\{-\lambda^2/O(\log n)\}, \quad (9)$$

as claimed.