

Chapter 9

Chaining methods

Much of this Chapter is unedited or incomplete.

1. The chaining strategy

The previous Chapter gave several ways to bound the tails of $\max_{i \leq N} |X_i|$. The chaining method applies such bounds recursively, taking advantage of extra structure on the index set.

Roughly speaking, the chaining method will work for any stochastic process $\{Z_t : t \in T\}$ for which we have some probabilistic control over the maxima of finite sets of increments $Z(s) - Z(t)$. For the basic arguments we may assume T is finite. It will be important that the bounds do not depend explicitly on the size of T if we wish to get inequalities for infinite T by passing to the limit in inequalities for finite subsets of T .

The chaining method works by linking together approximations to Z based on the values it takes on different finite subsets of T . Typically T is equipped with a metric $d(\cdot, \cdot)$ (or pseudometric) and the subsets are chosen as δ -nets for various δ .

<1> **Definition.** A finite subset T_δ of T is a δ -net if $\min_{s \in T_\delta} d(s, t) \leq \delta$ for each t in T . Equivalently, T is the union of the closed balls of radius δ with centers in T_δ . The smallest value of $\#T_\delta$ for all possible δ -nets is called the covering number, which denoted by $N(\delta)$, or $N(\delta, d, T)$ if there is any ambiguity over the choice of metric.

REMARKS.

- (i) In practice we do not need to know $N(\delta)$ exactly; an upper bound will suffice. In particular, we can often avoid messy details by choosing an upper bound that is continuous and strictly decreasing in δ . To avoid tedious qualifications, I will sometime call a subset T_δ a δ -net if $\#T_\delta$ is no larger than the upper bound on $N(\delta)$.
- (ii) Often T itself will be a subset of a larger metric space S . As stated, the definition of a δ -net for T does not allow centers to lie in $S \setminus T$. As shown by Section 3, the restriction has only a minor effect on applications.

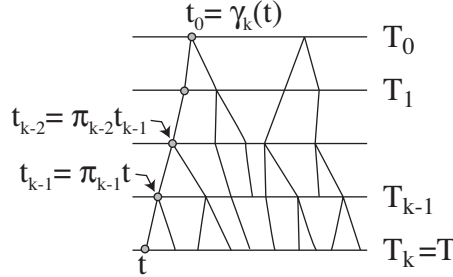
The typical chaining argument starts by choosing δ_i -nets T_i for numbers $\delta_0 > \delta_1 > \dots > \delta_k > 0$, with $T_k = T$. We then define π_i as the map that takes each t to its closest point in T_i , with some arbitrary rule for breaking ties. That

is, we construct maps $\pi_i : T \rightarrow T_i$ for which $d(t, \pi_i t) \leq \delta_i$. For the basic approximation argument we do not have to know that the maps π_i have been chosen in such a way, and we do not even have to require the T_i to be δ_i -nets.

<2> **Lemma.** Suppose $T_0, \dots, T_k = T$ are subsets of a finite set T . Suppose there exist maps $\pi_{i-1} : T_i \rightarrow T_{i-1}$. Define $\gamma_i : T_i \rightarrow T_0$ as the composition $\pi_{i-1} \circ \pi_{i-2} \circ \dots \circ \pi_0$. Then

$$\max_{t \in T} |Z(t) - Z(\gamma_k(t))| \leq M_1 + \dots + M_k$$

where $M_i := \max_{s \in T_i} |Z(s) - Z(\pi_{i-1}(s))|$.



Proof. Write D_i for $\max_{s \in T_i} |Z(s) - Z(\gamma_i(s))|$. Note that $D_1 = M_1$. For t in T write t_{k-1} for $\pi_{k-1}t$. Then

$$\begin{aligned} D_k &= \max_{t \in T_k} |Z(t) - Z(t_{k-1}) + Z(t_{k-1}) - Z(\gamma_{k-1}(t_{k-1}))| \\ &\leq \max_{t \in T_k} |Z(t) - Z(t_{k-1})| + \max_{t \in T_k} |Z(t_{k-1}) - Z(\gamma_{k-1}(t_{k-1}))| \\ &\leq M_k + D_{k-1} \end{aligned}$$

□ Argue similarly to bound D_{k-1} , and so on.

We could use the Lemma to bound the maximum of the process Z . For each t in T ,

$$Z_t \leq Z(\gamma_k(t)) + |Z_t - Z(\gamma_k(t))| \leq \max_{s \in T_0} Z_s + \max_{s \in T} |Z(s) - Z(\gamma_k(s))|.$$

Taking the maximum over t on the left-hand side we then get

$$<3> \quad \max_{t \in T} Z_t \leq \max_{s \in T_0} Z_s + \sum_{i=1}^k M_i$$

A very similar argument would show establish the analogous two-sided bound,

$$<4> \quad \max_{t \in T} |Z_t| \leq \max_{s \in T_0} |Z_s| + \sum_{i=1}^k M_i$$

Write N_i for $\#T_i$. Note that both the left-hand side of <3> and M_k involve a maximum over N_k variables. We can hope to get an improvement if the variables involved in M_k are “smaller than those involved in the left-hand side. It is here that control of the increments by a metric becomes important.

<5> **Example.** Let $\{Z_t : 0 \leq t \leq 1\}$ be a standard Brownian motion. We know that

$$\mathbb{P}\{\sup_t Z_t \geq x\} = 2\mathbb{P}\{Z_1 \geq x\} = \bar{\Phi}(x) \leq \frac{1}{2} \exp(-x^2/2)$$

Use Orlicz norm bound for $\psi(t) = \frac{1}{2} \exp(-t^2)$ to get comparable maximal inequality. Point out that the chaining method also works in higher dimensions.

□ More details needed.

2. Chaining inequalities for norms

In Chapter ??? we found several inequalities for maxima of finitely many random variables expressible in terms of norms. For example, if p th moments are finite then

$$\left\| \max_{i \leq N} |X_i| \right\|_p \leq N^{1/p} \max_{i \leq N} \|X_i\|_p$$

For an Orlicz norm defined by a convex increasing function ψ ,

$$\mathbb{P} \max_{i \leq N} |X_i| \leq \psi^{-1}(N) \max_{i \leq N} \|X_i\|_\psi$$

and

$$<6> \quad \mathbb{P}_A \max_{i \leq N} |X_i| \leq \psi^{-1}(N/\mathbb{P}A) \max_{i \leq N} \|X_i\|_\psi,$$

where \mathbb{P}_A denotes expectation conditional on an event A with $\mathbb{P}A > 0$. If the ψ function satisfies a **moderate growth condition**,

$$<7> \quad \psi(\alpha)\psi(\beta) \leq \psi(C_0\alpha\beta) \quad \text{for } \psi(\alpha) \wedge \psi(\beta) \geq 1,$$

where C_0 is a finite constant, then

$$<8> \quad \left\| \max_{i \leq N} |X_i| \right\|_\psi \leq C \psi^{-1}(N) \max_{i \leq N} \|X_i\|_\psi \quad \text{where } C := \frac{2 - \psi(0)}{1 - \psi(0)} C_0$$

For example, if $\psi(t) = \frac{1}{2} \exp(t^2)$ then condition <7> holds with $C_0 = 3/(\log 2)$, in which case $C = 9/(\log 2) \approx 13$.

The chaining method works well with any norm $\rho(\cdot)$ for random variables (such as an \mathcal{L}^p or Orlicz norm) for which there exists a (slowly) increasing function $H(\cdot)$ such that

$$<9> \quad \rho \left(\max_{i \leq N} |Z(s_i) - Z(t_i)| \right) \leq H(N) \max_{i \leq N} d(s_i, t_i)$$

REMARK. We do not need ρ to be a norm. It would suffice if it were a seminorm for which $\rho(X) = 0$ implies that $X = 0$ almost surely and $\rho(X) < \infty$ implies $|X| < \infty$ almost surely. If we work with equivalence classes of random variables for which $\rho(X) < \infty$ then we get a true norm. It is traditional to abuse notation and call a seminorm a norm.

We also need to assume that $\rho(X) \leq \rho(Y)$ whenever $|X| \leq |Y|$. Applying Lemma <2> with the T_i as δ_i -nets and the π_i as the maps to the nearest point of T_i , we then get

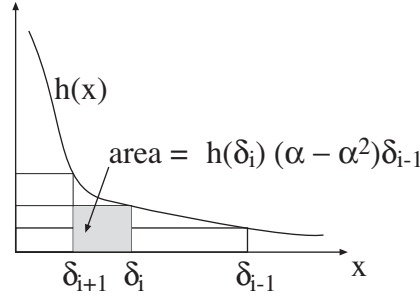
$$\begin{aligned} \rho \left(\max_{t \in T} |Z(t) - Z(\gamma_k(t))| \right) &\leq \rho(M_1) + \dots + \rho(M_k) \\ &\leq \sum_{i=1}^k H(N_i) \delta_{i-1} \end{aligned}$$

<10>

It is traditional to bound sums by integrals to make the inequalities look cleaner.

<11> **Lemma.** Let h be a nonnegative, decreasing function defined on an interval $(0, \delta]$. For a fixed α in $(0, 1)$, define $\delta_i := \delta\alpha^i$ for $i = 0, 1, 2, \dots$. Then

$$\sum_{i=1}^k \delta_{i-1} h(\delta_i) \leq \frac{1}{\alpha - \alpha^2} \int_{\delta_{k+1}}^{\delta_1} h(x) dx.$$



Proof. By monotonicity of h we have $(\delta_i - \delta_{i+1}) h(\delta_i) \leq \int_{\delta_{i+1}}^{\delta_i} h(x) dx$. Sum over i , noting that $\delta_i - \delta_{i+1} = \delta_{i-1}(\alpha - \alpha^2)$.

When we are not worried about precise values of constants it is often convenient to choose $\alpha = 1/2$ and expand the range of integration slightly, leaving a bound $2 \int_0^\delta h(x) dx$. Of course, if the integral is divergent we should not be so cavalier about a contribution from $(0, \delta_{k+1})$.

REMARK. A similar trick works if we partition the vertical axis in such a way that $h(\delta_i)$ increases geometrically fast. Some of the early papers in the empirical process literature used this variation of the method to bound sums by integrals.

To summarize, let me choose $\alpha = 2$ in Lemma <11> and also disguise the evidence of the chaining construction from Lemma <2> to get a neater result.

<12> **Theorem.** Let $\{Z_t : t \in T\}$ be a process indexed by a finite metric space T , with covering numbers $N(\cdot)$. Suppose $\rho(\cdot)$ is a norm on random variables for which $\rho(X) \leq \rho(Y)$ whenever $|X| \leq |Y|$. Suppose $H(\cdot)$ is an increasing function for which inequality <9> holds. Let T_δ be a δ -net for T . Then there is a map $\gamma : T \rightarrow T_\delta$ for which

$$(i) \quad d(t, \gamma(t)) \leq 2\delta \text{ for every } t \text{ in } T$$

$$(ii) \quad \rho\left(\max_{t \in T} |Z(t) - Z(\gamma(t))|\right) \leq 4 \int_{\delta/2^{k+1}}^{\delta/2} H(N(x)) dx, \text{ where } k \text{ is the smallest integer for which } \min\{d(s, t) : s \neq t\} \geq \delta/2^k.$$

Is this the right k ?

Proof. Invoke <10> for δ_i -nets T_i with $\delta_i = \delta/2^i$. Write γ instead of γ_k . Let the π_i 's map to the nearest point of T_i . For a given t in T , let $t = t_k \rightarrow t_{k-1} \rightarrow \dots \rightarrow t_1 \rightarrow t_0$ be the chain from t to $\gamma(t)$. Then

$$d(t, \gamma(t)) \leq d(t_k, t_{k-1}) + d(t_{k-1}, t_{k-2}) + \dots + d(t_1, t_0) \leq \delta_{k-1} + \delta_{k-2} + \dots + \delta_0 \leq 2\delta.$$

□ Invoke Lemma <11> to bound the sum from <10>.

<13> **Example.** Suppose the process $\{Z(t) : t \in T\}$ satisfies the bound

$$\|Z(s) - Z(t)\|_\psi \leq d(s, t) \quad \text{for all } s, t \in T,$$

with T a finite metric space, where the convex function ψ has the moderate growth property <7>.

Let $T_0 = \{t_0\}$ and $\delta := \max_{t \in T} d(t, t_0)$.

Apply Theorem <12> with ρ as the conditional \mathcal{L}^1 norm for \mathbb{P}_A and $H(N) = \psi^{-1}(N/\mathbb{P}_A)$ to get

$$<14> \quad \mathbb{P}_A X \leq 4 \int_0^\delta \psi^{-1}\left(\frac{N(x)}{\mathbb{P}_A}\right) dx \quad \text{where } X := \sup_t |Z(t) - Z(t_0)|.$$

The $N(x)$ can be disentangled from the $\mathbb{P}A$ using the moderate growth property for ψ . Invoke <7> with $x = \psi(\alpha)$ and $y = \psi(\beta)$ to deduce that

$$<15> \quad \psi^{-1}(xy) \leq C_0 \psi^{-1}(x) \psi^{-1}(y) \quad \text{for } x \wedge y \geq 1.$$

In particular,

$$\psi^{-1}\left(\frac{N(x)}{\mathbb{P}A}\right) \leq C_0 \psi^{-1}(N(x)) \psi^{-1}\left(\frac{1}{\mathbb{P}A}\right),$$

which, together with inequality <14>, implies

$$\mathbb{P}_A X \leq 4C_0 \psi^{-1}\left(\frac{1}{\mathbb{P}A}\right) \int_0^\delta \psi^{-1}(N(x)) dx.$$

Define $J := 4C_0 \int_0^\delta \psi^{-1}(N(x)) dx$. If we choose $A = \{X \geq \epsilon\}$, for a positive ϵ , then

$$\epsilon \leq \mathbb{P}_A X \leq \psi^{-1}\left(\frac{1}{\mathbb{P}A}\right) J,$$

from which it follows that

$$\mathbb{P}A \leq 1/\psi(\epsilon/J).$$

Compare with the tail bound we would get via a bound such as $\|X\|_\psi \leq J_0$.

You might find it enlightening to consult the book of Ledoux & Talagrand (1991), who have shown that the conditional \mathcal{L}^1 norm is ideally suited to

□ another, more powerful, method for deriving maximal inequalities.

3. Covering and packing numbers

Section not yet edited.

Suppose T is a set equipped with a pseudometric d . That is, d has all the properties of a metric except that distinct points might lie at zero distance. The slight increase in generality will allow us to equip function spaces with various \mathcal{L}^p norms (seminorms really) without too much fussing over almost sure equivalences.

For a subset A of T write $N_T(\delta, A, d)$ for the **δ -covering number**, the smallest number of closed δ -balls needed to cover A . That is, the covering number is the smallest N for which there exist points t_1, \dots, t_N in T with

$$\min_{i \leq N} d(t, t_i) \leq \delta \quad \text{for each } t \text{ in } A.$$

The set of centers $\{t_i\}$ is called a **δ -net** for A . Finiteness of all covering numbers is equivalent to total boundedness of A . Covering numbers are also called **metric entropies**.

Notice a small subtlety related to the subscript T in the definition. If we regard A as a pseudometric space in its own right, not just as a subset of T , then the covering numbers might be larger because the centers t_i would be forced to lie in A . It is an easy exercise (select a point of A from each covering ball that actually intersects A) to show that

$$N_A(2\delta, A, d) \leq N_T(\delta, A, d).$$

The extra factor of 2 will be of little consequence for the bounds derived in this Chapter. When in doubt, you should interpret covering numbers to refer to N_A .

On occasion it will prove slightly more convenient to work with the **packing number** $D(\delta, A, d)$, defined as the largest N for which there exist points t_1, \dots, t_N in A for which $d(t_i, t_j) > \delta$ if $i \neq j$. Notice the lack of a subscript T ; the packing numbers are an intrinsic property of A , and do not depend on T except through the pseudometric it defines on A . The $\delta/2$ -balls with centers at the t_i are disjoint; the balls are packed into A like oranges in a bag (perhaps protruding out into the larger space T).

<16> **Lemma.** For each $\delta > 0$,

$$N_A(\delta, A, d) \leq D(\delta, A, d) \leq N_T(\delta/2, A, d) \leq N_A(\delta/2, A, d).$$

Proof. For the middle inequality, observe that no closed ball of radius $\delta/2$ can contain points more than δ apart. Each of the centers for $D(\delta, A, d)$ must lie in a distinct $\delta/2$ covering ball. The other inequalities have similarly simple proofs. \square

I will refer to any calculation based on covering numbers or packing numbers as an **entropy method**, to avoid unfruitful distinctions.

<17> **Example.** Let T be the real line equipped with its usual metric d , and let $A = [0, 1]$. For $\delta < 1/2$, the $N + 1$ intervals of length 2δ and centers $\delta, 3\delta, \dots, (2N + 1)\delta, 1$ cover A if N is the largest integer such that $(2N + 1)\delta < 1 + \delta$. Thus $N_A(\delta, A, d) \leq \lceil (2\delta)^{-1} \rceil$. For a lower bound, note that the Lebesgue measure of the union of covering intervals of length 2δ must be no smaller than the Lebesgue measure of A . Thus $2\delta N_T(\delta, A, d) \geq 1$. The covering numbers increase like δ^{-1} as $\delta \rightarrow 0$. Actually, only the $O(\delta^{-1})$ upper bound will matter; \square the lower bound merely assures us that we have found the best rate.

<18> **Example.** Let $\|\cdot\|$ denote any norm on \mathbb{R}^k . For example, it might be ordinary Euclidean distance (the ℓ_2 norm), or the ℓ_1 norm, $\|x\|_1 = \sum_{i \leq k} |x_i|$. The covering numbers for any such norm share a common geometric bound.

Write B_R for the ball of radius R centered at the origin. For a fixed ϵ , with $0 < \epsilon \leq 1$, how many balls of radius ϵR does it take to cover B_R ? Equivalently, what are the packing numbers for B_R ?

Let x_1, \dots, x_N be a maximal set of points in B_R with $\|x_i - x_j\| > \epsilon R$ for $i \neq j$. The closed balls of radius $\epsilon R/2$ centered at the x_i are disjoint, and their union lies within $B_{R+\epsilon R/2}$. If we write Γ for the Lebesgue measure of the unit ball B_1 then

$$N(\epsilon R/2)^k \Gamma \leq (R + \epsilon R/2)^k \Gamma,$$

\square from which we deduce $N \leq ((2 + \epsilon)/\epsilon)^k \leq (3/\epsilon)^k$, for $0 < \epsilon \leq 1$.

4. Infinite index sets

Suppose the norm ρ from Theorem <12> also has the property

<19> if $0 \leq X_1 \leq X_2 \leq \dots \uparrow X$ then $\rho(X_n) \uparrow \rho(X)$.

Then we can pass to the limit in the inequality asserted by that Theorem to get bounds involving points from a countable dense subset of T . There are a few small subtleties in the construction, which I will illustrate by establishing a very useful equicontinuity bound.

<20> **Theorem.** Let $\{Z_t : t \in T\}$ be a process indexed by a metric space T , with covering numbers $N(\cdot)$. Suppose $\rho(\cdot)$ is a norm on random variables for which $\rho(X) \leq \rho(Y)$ whenever $|X| \leq |Y|$ and for which properties <9> and <19> hold. Suppose $H(\cdot)$ is an increasing function for which inequality <9> holds and for which

$$<21> \quad \int_0^1 H(N(x)) dx < \infty$$

Then:

(i) There exists a countable dense subset T_∞ of T for which: to each $\epsilon > 0$ there exists an $\eta > 0$ such that

$$\rho \left(\sup |Z_s - Z_t| : s, t \in T_\infty \text{ and } d(s, t) < \eta \right) \leq \epsilon$$

(ii) Almost all sample paths of $\{Z_t : t \in T_\infty\}$ are uniformly continuous.

(iii) There exists a process $\{\tilde{Z}_t : t \in T\}$ with uniformly continuous sample paths such that $\mathbb{P}\{\tilde{Z}_t = Z_t\} = 1$ for each t in T and for which

$$\rho \left(\sup |\tilde{Z}_s - \tilde{Z}_t| : s, t \in T \text{ and } d(s, t) < \eta \right) \leq \epsilon$$

Proof. It will be easier to work with packing numbers $D(\cdot)$ rather than covering numbers. The finiteness condition <21> still holds if we replace $N(x)$ by $D(x)$ because (Lemma <16>) $D(x) \leq N(x/2)$. Choose a $\delta > 0$ for which

$$\int_0^\delta H(D(x)) dx \leq \epsilon$$

Define $\delta_i := \delta/2^i$ for $i = 0, 1, 2, \dots$. Construct sets $T_0 \subseteq T_1 \subseteq T_2 \subseteq \dots$ by choosing T_0 as a maximal set of points for which $d(s, t) > \delta_0$ if $s \neq t$, for $s, t \in T_0$. Then add extra points to T_0 to create a maximal set of points $T_1 \supseteq T_0$ for which $d(s, t) > \delta_1$ if $s \neq t$, for $s, t \in T_1$. And so on. Thus $\#T_k \leq D(\delta_k)$ for each k and T_k is a δ_k -net for T . Moreover,

$$T_k \uparrow T_\infty := \cup_i T_i \quad \text{as } k \uparrow \infty.$$

Construct chains and maps $\gamma_k : T_k \rightarrow T_0$ as in Section 1.

Temporarily hold k fixed. Invoke Theorem <12> to show that

$$\rho(G_k) \leq 4 \int_0^\delta H(N(x)) dx \leq 4\epsilon \quad \text{where } G_k := \max_{t \in T_k} |Z(t) - Z(\gamma_k(t))|.$$

Now we come to a subtle part of the argument, making use of a clever construction from Ledoux & Talagrand (1991, Section 11.1).

The map γ_k partitions T_k into $N \leq \#T_0 \leq D(\delta)$ equivalence classes E_1, \dots, E_N , by means of the relation $s \sim t$ if $\gamma_k s = \gamma_k t$. If $s \sim t$ then

$$<22> \quad |Z(s) - Z(t)| \leq |Z(s) - Z(\gamma_k s)| + |Z(\gamma_k t) - Z(t)| \leq 2G_k.$$

For an as yet unspecified $\eta > 0$, write $E_i \approx E_j$ if there exist points $t_{ij} \in E_i$ and $t_{ji} \in E_j$ such that $d(t_{ij}, t_{ji}) < \eta$. Define

$$G := \max_{E_i \approx E_j} |Z(t_{ij}) - Z(t_{ji})|.$$

The maximum runs over at most N^2 pairs (t_{ij}, t_{ji}) . By inequality <9>

$$\rho(G) \leq H(N^2)\eta \leq H(D(\delta)^2)\eta,$$

which is less than ϵ if η is chosen small enough.

If $S \subseteq T$ define

$$M(S, \eta) := \sup\{|Z_s - Z_t| : d(s, t) < \eta \text{ and } s, t \in S\}.$$

Of course, if S is finite then the sup could be replaced by a max.

Any difficulty if $i = j$?

A multi-step approximation will let us reduce comparison of pairs in T_k to comparison between pairs in T_0 . Suppose s and t are points of T_k such that $d(s, t) < \eta$. If $s \in E_i$ and $t \in E_j$ then $E_i \approx E_j$, and $s \sim t_{ij}$ and $t \sim t_{ji}$. It follows that

$$\begin{aligned} |Z(s) - Z(t)| &\leq |Z(s) - Z(t_{ij})| + |Z(t_{ij}) - Z(t_{ji})| + |Z(t_{ji}) - Z(t)| \\ &\leq 2G_k + G + 2G_k. \end{aligned}$$

Take the maximum over all such (s, t) pairs then take norms of both sides.

$$<23> \quad \rho(M(T_k, \eta)) \leq 4\rho(G_k) + \rho(G) \leq 5\epsilon.$$

You should repeat the argument with ϵ replaced by $\epsilon/5$ if you want the final inequality to exactly as stated.

Now let k tend to infinity. Each pair (s, t) that contributes to the supremum in $M(T_\infty, \eta)$ must appear in some T_k . It follows that

$$M(T_k, \eta) \uparrow M(T_\infty, \eta)$$

Invoke property <19> of the norm ρ to deduce assertion (i) of the Theorem.

To show that almost all sample paths of $\{Z_t : t \in T_\infty\}$ are uniformly continuous, invoke (i) to find a sequence $\{\eta_m\}$ for which $M(T_\infty, \eta_m) \leq 2^{-m}$. Then, by the continuity property <19>,

$$\rho\left(\sum_{m \in \mathbb{N}} M(T_\infty, \eta_m)\right) \leq \sum_{m \in \mathbb{N}} \rho(M(T_\infty, \eta_m)) < \infty.$$

$$\rho\left(\sum_{m \in \mathbb{N}} M(T_\infty, \eta_m)\right) \leq \sum_{m \in \mathbb{N}} \rho(M(T_\infty, \eta_m)) < \infty.$$

The sum $\sum_{m \in \mathbb{N}} M(T_\infty, \eta_m)$ is finitely almost surely and, consequently there exists a negligible set \mathcal{N} such that

$$M(T_\infty, \eta_m) \rightarrow 0 \quad \text{for } \omega \in \mathcal{N}^c.$$

The sample paths for $\omega \in \mathcal{N}^c$ are uniformly continuous (as a function on T_∞). For those ω , the path extends to a unique uniformly continuous function $\tilde{Z}_t(\omega)$ on T . Define $\tilde{Z}_t(\omega) \equiv 0$ for $\omega \in \mathcal{N}$.

Finish the argument

For pairs s, t in T for which $d(s, t) < \eta$, find sequences $\{t_k\}$ and $\{s_k\}$ in T_∞ for which $d(s_k, s) \rightarrow 0$ and $d(t_k, t) \rightarrow 0$. Then what?

□

5. Chaining with random distances

Section not yet edited. Please ignore.

<24> **Theorem.** Let $\{\Delta(s, t) : (s, t) \in T \times T\}$ be a random distance with $\Delta(s, t) \in \mathcal{M}_\rho^+$ for all (s, t) . Suppose there exists an increasing function H for which

$$\rho\left(\max_{i=1}^N \Delta(s_i, t_i)\right) \leq H(N) \max_{i=1}^N d(s_i, t_i)$$

for all finite sets of pairs $(s_1, t_1), \dots, (s_N, t_N)$. Then for each δ -net T_δ and each finite subset S of $T \dots$

$$\rho\left(\max_{t \in S} \Delta(t, \gamma_t)\right) \leq 4 \int_0^\delta H(N(x)) dx.$$

Proof.

$$\begin{aligned} \rho\left(\max_{t \in S} \Delta(t, \gamma_t)\right) &\leq \rho\left(\max_{t \in S} (\Delta(t, t_k)) + \sum_{j=1}^k \rho(M_j)\right) \\ &\leq H(\#S) \left(2 \sum_{j \geq k} \delta_j\right) + \sum_{j=1}^k H(N(\delta_j)) \delta_{j-1} \end{aligned}$$

□

<25> **Example.** Suppose $\{Z(t) : t \in [0, 1]\}$ is a process for which there exists a finite measure μ on $[0, 1]$ and constants $\gamma > 0$ and $\alpha > 1$ for which

$$(\mathbb{P}|Z(s) - Z(t)|^\gamma)^{1/\alpha} \leq \mu(s, t),$$

for all $0 \leq s < t \leq 1$. Show that Z has a version with cts paths. Argue first with $\gamma \geq \alpha$, then the other case, for $0 < \gamma < \alpha$, as in L&T. [Comment on the usefulness of the two-parameter process in the chaining argument.]

cf. Ledoux & Talagrand (1991, □
page 308)

6. Maximal inequalities for tail probabilities

Section not yet edited. Please ignore.

Let $\{\Delta(s, t) : (s, t) \in T \times T\}$ be a random distance indexed by a pseudometric space (T, d) for which we have a bound $N(\cdot)$ on the covering numbers. Suppose the tail probabilities for $\Delta(s, s')$ are controlled by the pseudometric by means of a nonnegative function $\beta(\cdot, \cdot)$, which is decreasing in its first argument and increasing in its second argument, such that

$$<26> \quad \mathbb{P}\{\Delta(s, s') \geq \eta\} \leq \beta(\eta, d(s, s')) \quad \text{for } s, s' \in T \text{ and } \eta \geq 0.$$

For N pairs (s_i, s'_i) each with $d(s_i, s'_i) \leq \delta$ we then have a bound,

$$<27> \quad \mathbb{P}\{\max_{i \leq N} \Delta(s_i, s'_i) \geq \eta\} \leq N\beta(\eta, \delta).$$

For nonnegative numbers $\eta, \eta_1, \eta_2, \dots$, Lemma <GENERAL.CHAIN> gives, for each finite subset S of T , a maximal inequality:

$$\begin{aligned} &\mathbb{P}\{\max_{t \in S} \Delta(t, \gamma_t) \geq \eta + \eta_1 + \dots + \eta_k\} \\ &\leq \mathbb{P}\{\max_{t \in S} \Delta(t, t_k) \geq \eta\} + \sum_{i=1}^k \mathbb{P}\{M_i \geq \eta_i\} \\ &\leq (\#S) \beta(\eta, \delta_k + \sum_{i \geq k} \delta_i) + \sum_{i=1}^k N(\delta_i) \beta(\eta_i, \delta_{i-1}). \end{aligned}$$

Provided $\beta(\eta, \delta) \rightarrow 0$ as $\delta \rightarrow 0$ for each fixed η , we also have a limiting form of the maximal inequality:

$$<29> \quad \mathbb{P}\{\max_{t \in S} \Delta(t, \gamma_t) > \sum_{i=1}^{\infty} \eta_i\} \leq \sum_{i=1}^{\infty} N(\delta_i) \beta(\eta_i, \delta_{i-1}).$$

Notice the strict inequality on the left-hand side, to accommodate a small positive η .

If the covering bound $N(x)$ increases slowly enough as x tends to zero, and if $\beta(\eta, \delta)$ tends to zero rapidly enough when $\eta \rightarrow \infty$ and $\delta \rightarrow 0$ at appropriate rates, the maximal inequalities can be expressed in slightly more explicit forms. It is traditional to bound sums by integrals to make the inequalities look even simpler.

<30> **Example.** Let $\{Z_t : t \in T\}$ be a stochastic process whose increments satisfy a subgaussian inequality controlled by the pseudometric on T :

$$<31> \quad \mathbb{P}\{|Z_s - Z_t| \geq \eta d(s, t)\} \leq c_0 \exp(-c_1^2 \eta^2),$$

for some positive constants c_0 and c_1 . That is, the tail bound <26> holds with $\beta(\eta, \delta) = c_0 \exp(-c_1^2 \eta^2 / \delta^2)$. Once again write $\Delta(s, t)$ for $|Z_s - Z_t|$. Inequality <29> becomes

$$\mathbb{P}\{\max_{t \in S} \Delta(t, \gamma_t) > \sum_{i=1}^{\infty} \eta_i\} \leq c_0 \sum_{i=1}^{\infty} N(\delta_i) \exp(-c_1^2 \eta_i^2 / \delta_{i-1}^2).$$

We need to choose the $\{\eta_i\}$ to make the sum on the right-hand side converge. A geoemtric rate of decrease would ensure that the sum behaves like its first term. For a fixed, positive x define η_i so that

$$\exp(-c_1^2 \eta_i^2 / \delta_{i-1}^2) = e^{-x} 2^{-i} / N(\delta_i),$$

that is,

$$\begin{aligned} \eta_i &:= c_1^{-1} \delta_{i-1} \sqrt{\log N(\delta_i) + i \log 2 + x} \\ &\leq c_1^{-1} 2\delta_i \left(h(\delta_i) + \sqrt{i \log 2} + \sqrt{x} \right) \quad \text{where } h(y) := \sqrt{\log N(y)}. \end{aligned}$$

With the help of Lemma <11> we then get

$$\sum_{i=1}^{\infty} \eta_i \leq c_2 \int_0^{\delta/2} h(y) dy + c_3 \delta + c_4 \delta \sqrt{x},$$

where $c_2 := 4c_1^{-1}$ and $c_3 := 2c_1^{-1} \sqrt{\log 2} \sum_{i=1}^{\infty} \sqrt{i}/2^i$ and $c_4 := 2c_1^{-1}$. Assume that the covering bounds increase slowly enough that the integral

$$J_z := \int_0^z \sqrt{\log N(y)} dy$$

is convergent for each $z > 0$. Then, for each finite subset S of T ,

$$<32> \quad \mathbb{P}\{\max_{t \in S} |Z_t - Z_{\gamma_t}| > c_2 J_\delta + c_3 \delta + c_4 \delta \sqrt{x}\} \leq c_0 e^{-x},$$

where the constants c_3, c_4 , and c_5 depend only on the c_1 from <31>.

Now suppose that T has radius at most R , in the sense that there is some point τ in T for which $\sup_{t \in T} d(t, \tau) = R < \infty$, and that we wish to determine how large a value w is needed to make the tail probability $\mathbb{P}\{\max_{t \in S} |Z_t - Z_\tau| > w\}$ smaller than a prescribed quantity, which for convenience I write as $2c_0 e^{-x}$.

As t ranges over S , the value γ_t ranges over a subset of the δ -net T_δ , a set with at most $N(\delta)$ points each at a distance at most R from τ . The inequality

$$\max_{t \in S} |Z_t - Z_\tau| \leq \max_{t \in T_\delta} |Z_t - Z_\tau| + \max_{t \in S} |Z_t - Z_{\gamma_t}|$$

then leads us to a bound

$$\mathbb{P}\{\max_{t \in S} |Z_t - Z_\tau| > w + c_2 J_\delta + c_3 \delta + c_4 \delta \sqrt{x}\} \leq N(\delta) c_0 \exp(-c_1^2 w^2 / R^2) + c_0 e^{-x}.$$

REMARK. Notice that I have built in the assumption that a reasonable way to make a sum of two terms small is to put each of them equal to half the desired sum. Perhaps a significantly better bound could be obtained by discarding the assumption and trying to optimize over the allocation of how much of the final tail bound comes from each term.

For a given x we are left with the task of choosing δ and w to make

$$w + c_2 J_\delta + c_3 \delta + c_4 \delta \sqrt{x} \quad \text{small subject to } N(\delta) \exp(-c_1^2 w^2 / R^2) \leq e^{-x}.$$

Of course there is no point in making δ larger than R , because we may assume $N(y) = 1$ for $y > R$. Also, we may suppose x is bounded away from zero

(say $x > c_5$), because there is no point in trying to optimize when $c_0 e^{-x}$ is not a lot smaller than 1.

The smallest w satisfying the constraint is $(R/c_1)\sqrt{x + h(\delta)^2}$, where once again $h(y) = \sqrt{\log N(y)}$. We have the formidable task of finding $\delta \in (0, R]$ to minimize

$$\frac{R}{c_1} \sqrt{x + h(\delta)^2} + c_2 \int_0^\delta h(y) dy + \delta(c_3 + c_4 \sqrt{x}).$$

□

7. Chaining via majorizing measures

Talagrand (1996)

8. Problems

More problems to come

- [1] Show that the convex function $\psi(x) = (1+x) \log(1+x) - x$ does not satisfy the growth condition $\langle 7 \rangle$. Hint: Consider the limit of $\psi(C_0 x^2)/\psi(x)^2$ as $x \rightarrow \infty$.

9. Notes (inaccurate and incomplete)

Acknowledge Ledoux & Talagrand (1991) for several of the ideas used in this Chapter: Example $\langle \text{CONDIT.MEAN} \rangle$; the introduction of the two-parameter process in the proof of Theorem $\langle \text{FINITE.MAXIMAL} \rangle$ (and its usefulness in the analog of Example $\langle 25 \rangle$ for $0 < \gamma < \alpha$); Example $\langle \text{ORLICZ2} \rangle$, or maybe cite Pisier; the subtle equivalence class idea in Example $\langle \text{ORLICZ2} \rangle$; and the method used in Example $\langle 13 \rangle$.

Give some history of earlier work: Dudley, Pisier?
van der Vaart & Wellner (1996)

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