

Math 254B Lecture 27 Notes

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1 Ergodic Decomposition of CP Distributions and Typical Dimension Under Ergodic CP Distributions

1.1 Lemma for obtaining an adapted T -invariant distribution

Recall: We have $K_w = \phi_w[K]$, where $w \in [k]^n$. Define $\nu^w := S_*^n \nu|_{K_w}$. If $z \in \text{supp}(\nu)$, then $T^t(z, \nu) = (S^t z, \nu^{\alpha_{[1,t]}(z)})$.

$\hat{\mu}$ on $K \times P(K)$ is adapted if

$$\hat{\mu} = \int_{P(K)} \nu \times \delta_\nu d\bar{\mu}(\nu) \iff \int f d\hat{\mu} = \int Qf d\bar{\mu}, \quad Qf(z, \nu) = \int f(z', \nu) d\nu(z').$$

We denote P_a as the collection of adapted distributions in $P(K \times P(K))$.

We need to prove the following lemma to complete our proof that we can still run the Krylov-Bogoliubov machine.

Lemma 1.1. *If $\hat{\mu} \in P_a$, then $T_*\hat{\mu} \in P_a$, and its second marginal is*

$$M\bar{\mu} = \int \sum_{i=1}^k \nu(K_i) \cdot \delta_{\nu^i} d\bar{\mu}(\nu).$$

Proof. We have

$$\begin{aligned} \int f \circ T d\hat{\mu} &= \int f(Sz, \nu^{\alpha_1(z)}) d\hat{\mu}(z, \nu) \\ &= \iint f(Sz, \nu^{\alpha_1(z)}) \nu(z) d\bar{\mu}(\nu) \\ &= \int \sum_{i=1}^k \nu(K_i) \int_{K_1} f(Sz, \nu^i) d\nu|_{K_i}(z) d\bar{\mu}(\nu) \\ &= \int \sum_{i=1}^k \nu(K_i) \int_{K_1} f(Sz, \nu^i) d \underbrace{S_*\nu|_{K_i}(z)}_{\nu^i(z)} d\bar{\mu}(\nu) \end{aligned}$$

$$= \iint f(z, \nu') d\nu'(z) d(M\bar{\mu})(\nu'). \quad \square$$

1.2 Ergodic decomposition of CP distributions

Definition 1.1. A **CP distribution** is an adapted and T -invariant distribution.

Proposition 1.1. If $\hat{\mu}$ is a CP distribution with ergodic decomposition $\hat{\mu} = \int_X \hat{\mu}_x d\hat{\mu}(x)$, then $\hat{\mu}$ -a.e. x is adapted (and hence CP).

This is a bit technical, so we need the following lemma.

Lemma 1.2. If $\hat{\mu}$ is CP and $f \in L^1(\hat{\mu})$, then

$$\lim_{n \rightarrow \infty} A_n f = \lim_{n \rightarrow \infty} A_n(Qf) \quad \hat{\mu}\text{-a.e.}$$

Proof. By the ergodic theorem, the limits exist.

Step 1: It is enough to show that $(f - Qf) \rightarrow 0$ in $\|\cdot\|_1$ for all $f \in C(K \times P(K))$.

Step 2: It is enough to show that $A_n f \rightarrow 0$ in $\|\cdot\|_1$ when f is continuous and $Qf = 0$.

Step 3: It is enough to show that $\int g \cdot A_n f d\hat{\mu} \rightarrow 0$ for all $g \in C(K \times P(K))$ if $Qf = 0$.

Step 4: By Stone-Weierstrass, it is enough to test when $g(z, \nu) = g_1(\alpha_{[1,m]}(z))g_2(\nu)$.

We can now compute

$$\int f \cdot (f \circ T^t) d\hat{\mu} = \int_{P(k)} \left[\int_K g_1(\alpha_{[1,m]}(z)) f(S^t z, \nu^{\alpha_{[1,t]}(z)}) d\nu(z) \right] g_2(\nu) \bar{\mu}(z)$$

If $t > m$,

$$= \int_{P(k)} \left[\int_K g_1(\alpha_{[1,m]}(z)) \underbrace{\mathbb{E}[f(S^t(\cdot), \nu^{\alpha_{[1,t]}(z)}) \mid \alpha_{[1,t]}](z)}_{\mathbb{E}_\nu[f(S^t(\cdot), \nu^w) \mid K_w]} d\nu(z) \right] g_2(\nu) \bar{\mu}(z).$$

Observe that

$$\mathbb{E}_\nu[f(S^t(\cdot), \nu^w) \mid K_w] = \int f(z', \nu^w) dS_{*|K_w}^t(z') = \int f(z', \nu^w) \nu^2(z') = Qf(z, \nu^w) = 0.$$

So for $t > m$, the quantity is 0. So

$$\int g \cdot A_n f d\hat{\mu} = \frac{1}{n} \sum_{t=0}^{n-1} \int g \cdot (f \circ T^t) d\hat{\mu} \rightarrow 0, \quad \square$$

We can now prove the proposition:

Proof. Let $f \in C(L \times P(K))$. Then

$$\widehat{\mu}_x(\{(z, \nu) : \lim_n A_n f(z, \nu) = \int f d\widehat{\mu}_x\}) = 1,$$

$$\widehat{\mu}_x(\{(z, \nu) : \lim_n A_n Qf(z, \nu) = \int Qf d\widehat{\mu}_x\}) = 1.$$

So $\int f d\widehat{\mu}_x = \int Qf d\widehat{\mu}_x$ for $\widehat{\mu}$ -a.e. x . □

1.3 Typical dimension under ergodic CP distributions

Convention: IF (X, \mathcal{B}, μ) is a probability space, “ μ -a.e. x satisfies P ” means “there exists $A \in \mathcal{B}$ such that $\mu(A) = 1$ and x satisfies P for all $x \in A$.” This allows us to not check if sets are measurable.¹

Proposition 1.2. *If $\widehat{\mu}$ is ergodic and CP, then $\widehat{\mu}$ -a.e. (z, ν) satisfies*

$$\dim(\nu) = \frac{\int H_{\nu'}(\alpha_1) \widehat{\mu}(z', \nu')}{\log(r^{-1})}.$$

Proof. Define $F(z, \nu) = -\log(\nu([z]_1))$. Then

$$\int F \widehat{\mu} = \int_{P(K)} \underbrace{\int_K -\log(\nu([z]_1)) d\nu(z)}_{-\sum_{i=1}^k \nu(K_i) \log(\nu(K_i))} d\widehat{\mu}(\nu) = \int H_{\nu}(\alpha) d\widehat{\mu}(z', \nu').$$

By the pointwise ergodic theorem for $\widehat{\mu}$ -a.e. (x, ν) ,

$$A_n F(z, \nu) = -\frac{1}{n} \sum_{t=0}^{n-1} \log(\nu|_{[z]_1^t}([z]_{t+1})) - \frac{1}{n} \log(\nu([z]_1^n)) \rightarrow \int F d\widehat{\mu}.$$

In other words,

$$\nu([z]_1^n) = e^{-hn+o(n)}, \quad h = \int H_{\nu'}(\alpha_1) d\widehat{\mu}(z', \nu').$$

as $n \rightarrow \infty$. So

$$\dim(\nu, z) = \frac{h}{\log(r^{-1})}$$

for $\widehat{\mu}$ -a.e. (z, ν) . Since $\widehat{\mu}$ is adapted,

$$\dim(\nu, z) = \frac{h}{\log(r^{-1})}$$

(for ν -a.e. z) for $\widehat{\mu}$ -a.e. ν . □

¹A projection of a Borel subset of $X \times Y$ need not be Borel.