

High entropy method

G \mathbb{R} -points of a linear \mathbb{R} -group

ACG \mathbb{R} -points of a \mathbb{R} -split torus of rank $r \geq 2$

$I < G$ discrete subgroup, $X = G/I$

μ Borel A -invariant A -ergodic probability measure on X

Weight decomposition of $\mathfrak{g} = \text{Lie}(G)$

$$\mathfrak{g} = \bigoplus_{\lambda \in \Phi} \mathfrak{g}^\lambda \quad \Phi \subset \{\text{character of } A\} \text{ is the set of weights}$$

$$\mathfrak{g}^\lambda = \{x \in \mathfrak{g} \mid \forall a \in A, \text{Ad}(a)x = \lambda(a)x\}$$

Observation $[\mathfrak{g}^\lambda, \mathfrak{g}^\eta] \subset \mathfrak{g}^{\lambda+\eta} \quad \forall \lambda, \eta \in \Phi$

Proof $\forall a \in A \quad A \text{Ad}(a)[X, Y] = [\text{Ad}(a)X, \text{Ad}(a)Y]$
 $\forall X, Y \in \mathfrak{g}$

Notation for $\Theta \subset \Phi$, write

$$\mathfrak{g}^\Theta = \bigoplus_{\lambda \in \Theta} \mathfrak{g}^\lambda$$

For example $\text{Lie}(G_a^-) = \mathfrak{g}^{\{\lambda \in \Phi \mid |\lambda(a)| < 1\}} = \Phi \cap \left(\text{an open half space in } \mathfrak{g}^{\mathbb{R}^r} \right)$

Definition $\lambda \sim \eta$ if $\lambda^n = \eta^m$ for some $n, m \geq 1$,

and $[\lambda] = \{y \in \Phi \mid y \sim \lambda\}$

$\mathfrak{g}^{[\lambda]}$ is called a **Coarse Lyapunov subalgebra**.

Theorem Let $\lambda, \eta \in \Phi$ with $[\lambda] \neq [\eta] \neq [\lambda']$

μ is invariant under $[H_x^{[\lambda]}, H_x^{[\eta]}]$ for $a \in x \in X$

where $H_x^{[\lambda]}$ is the smallest A -normalised connected closed subgroup containing $\text{Supp}(\mu_x^{G^{[\lambda]}})$.

For subgroup $U \subset G_a^-$ for some $a \in A$, for $x \in X$

$$\text{St}_x^U = \text{Stab}_x \mu_x^U = \{u \in U \mid u \cdot \mu_x^U = \mu_x^U\}$$

Lemma $\forall a \text{ a.e. } x \in X, \quad U \subset \bar{G}_a \text{ normalised by } A$

(1) $St_x^u \subset Stab_G(a)$

(2) $\forall a \in A \quad a St_x^u a^{-1} = St_{a \cdot x}^u$

(3) $\Leftrightarrow Ad(a) St_x^u = St_{a \cdot x}^u$

(3) St_x^u is closed and connected.

(4) St_x^u is normalised by A .

Notation $st_x^u = Lie(St_x^u)$

(2) + (4) + ergodicity \Rightarrow St_x^u same for a.e. x .

Proof (1) \checkmark

(2) $\Leftrightarrow \mu_{a \cdot x}^u = a^{-1} \cdot \mu_x^u \cdot a$

(3) closed. $(u_n \rightarrow u, u_n \cdot \mu_x^u \rightarrow^* u \cdot \mu_x^u)$

let $f(x) = d(St_x^u)^0, (St_x^u) \setminus (St_x^u)^0$

by (2) $f(a^n x) \rightarrow 0$
 $U \subset \bar{G}_a$

By Poincaré recurrence $\mu(\{x \in X \mid f(x) > 0\}) = 0$.

(4) $\Leftrightarrow Ad(a) St_x^u = St_x^u \quad \forall a \in A$.

By ergodicity + (2), $\dim St_x^u = k$ is the same.

$A \curvearrowright \Lambda^k \mathfrak{g}$

$\rho(a) = \Lambda^k Ad(a)$

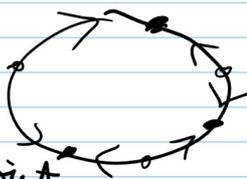
This is algebraic. By the fact (last week), $\rho(A)$ is diagonalisable

$\forall a \in A, \rho(a) \sim P(\Lambda^k \mathfrak{g})$

$\forall \theta \in P(\Lambda^k \mathfrak{g})$

either $\rho(a)\theta = \theta$

or $\rho(a^n)\theta \rightarrow$ a fixed point



Apply to $\theta = \Lambda^k St_x^u \in P(\Lambda^k \mathfrak{g})$

θ not fixed point contradicts Poincaré recurrence.

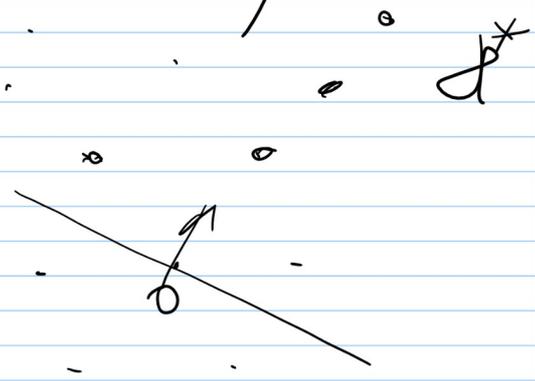
Hence $Ad(a) St_x^u = St_x^u$.



Rmk $\mathfrak{g} \subset \mathfrak{g}^*$ is normalised by A .

then $(\forall x \in \mathfrak{g}, x = \sum_{\lambda \in \Phi} x_\lambda, x_\lambda \in \mathfrak{g}^{\lambda})$
 then $x_\lambda \in \mathfrak{g} \quad \forall \lambda \in \Phi$.

\mathfrak{g} is A -norm. $\Leftrightarrow \mathfrak{g} = \bigoplus_{\lambda \in \Phi} (\mathfrak{h} \cap \mathfrak{g}^{\lambda})$



Proof of theorem

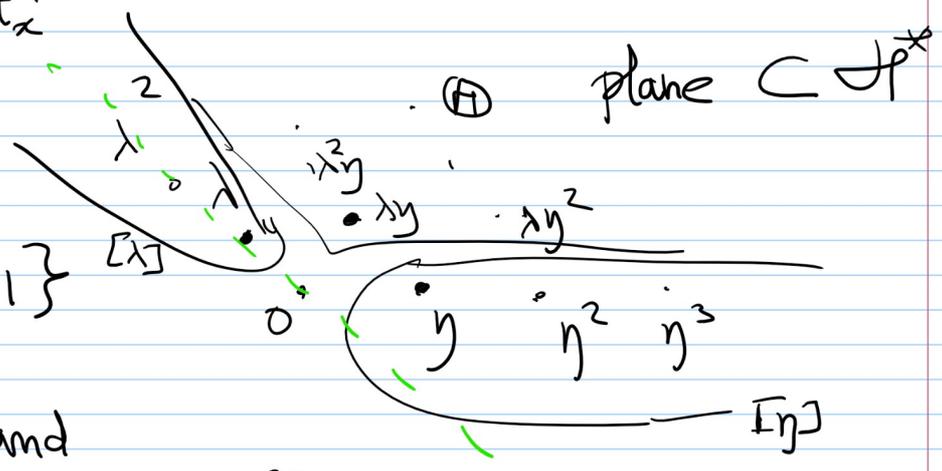
Step 0 $\forall y \in \text{supp } \mu_x \quad \forall h \in \text{supp } \mu_x$

then $(g, h) \in \text{St}_x^u$

where

$U = G^{\oplus 1}$

$\oplus = \{ \lambda y^m \mid m \geq 1 \}$



$\exists a \in A, \lambda(a) = 1$ and $S(a) < 1 \quad \forall S \in \oplus \cup [y]$

$G^{\oplus \cup [y]} \subset G_a^-$

and G^{\oplus} is centralised by a

and $[g^{\oplus}, g^{\oplus \cup [y]}] \subset g^{\oplus \cup [y]}$

hence $G^{\oplus \cup [y]}$ is normalised by G^{\oplus} .

By product theorem

$\mu_x \times \mu_x \cong (\mu_x \times \mu_x)$

$c: G \times G \rightarrow G, (x, y) \mapsto xy$

Similarly $\mu_x \times \mu_x \cong (\mu_x \times \mu_x)$

So

$$\mu_x^u \propto \mathcal{L}_* \left(\mu_x^{[\lambda]} \times \mu_x^{[\eta]} \times \mu_x^u \right)$$

$$\propto \mathcal{L}_* \left(\mu_x^{[\eta]} \times \mu_x^{[\lambda]} \times \mu_x^u \right)$$

Then by Fubini:

$$\mu_x^u \text{ is inv by } (g, h) = ghg^{-1}h^{-1} \text{ for}$$

$$\mu_x^{[\lambda]} \text{ a.e. } g \text{ and } \mu_x^{[\eta]} \text{ a.e. } h.$$

Step 1. If $g \in \text{supp}(\mu_x^{[\lambda]})$, $h \in \text{supp}(\mu_x^{[\eta]})$

$$\text{and } \log g = u = u_1 + u_2 + \dots$$

$$\uparrow \uparrow$$

$$\mathfrak{p}^{\lambda} \quad \mathfrak{p}^{\lambda^2}$$

$$\log h = v = v_1 + v_2 + \dots$$

$$\uparrow \uparrow$$

$$\mathfrak{p}^{\eta} \quad \mathfrak{p}^{\eta^2}$$

Byk
 $\{ (g, h) \}$
 $ghg^{-1}h^{-1} \in \text{supp}(\mu_x^u)$
 is closed

then $[u, v] \in \text{st}_x^u$.

Proof of Step 1. $\log(ghg^{-1}h^{-1}) \in \text{st}_x^u$

|| - Baker-Campbell-Hausdorff

$[u, v] + \text{terms of more brackets in } u, v$

$$= [u, v] + \text{terms in } \mathfrak{p}^{(\lambda)} \setminus \{ \lambda \}$$

$$\mathfrak{p}^{(\lambda)} = \mathfrak{p} \cup [\lambda] \cup [\eta]$$

st_x^u is \mathfrak{A} -normalised

$$\Rightarrow [u, v] \in \text{st}_x^u$$



Step 2 $\forall n, m \geq 1$. $[u_n, v_m] \in \text{st}_x^u$

Why do we care?

$$\mathfrak{g}^{[X]} = \text{Lie}(H_x^{[X]})$$

$$U_n = P_x^n(\log g)$$

$$U_m = P_y^m(\log h)$$

= Lie (smallest closed connected A -normalised subg. containing $\text{supp } \mu_x^{[X]}$)

= Lie algebra generated by $\bigcup_{n \geq 1} P_x^n(\text{supp } \mu_x^{[X]})$

$P_x^n: \mathfrak{g} \rightarrow \mathfrak{g}^{\otimes n}$ projection // $\forall \mathfrak{g} \subseteq \mathfrak{sl}(n, \mathbb{R})$

Step 3 $[\mathfrak{g}^{[X]}, \mathfrak{g}^{[Y]}] \subset \text{st}^u$

Step 4 $[H^{[X]}, H^{[Y]}] \subset \text{st}^u$

\downarrow BCH formula

Proof of step 2

Induction on the size of \mathbb{H}

($\mathbb{H} = \emptyset$)
Trivial

Induction on (n, m) (lexicographical order)

$(1, 1)$ was step 1.

Not show $[u_2, v_1] \in \text{st}^u$

$\exp([u_2, v_1]) \in \text{St}^u$ by step 1.

implies

$\in \text{supp } \mu_x^u$

$\in \text{supp } \mu_x^{[Y]}$

(tower rule)

by induction hypothesis (outer)

$[u_2, v_1], u_n \in \text{st}^u$

$\forall n \geq 1$.

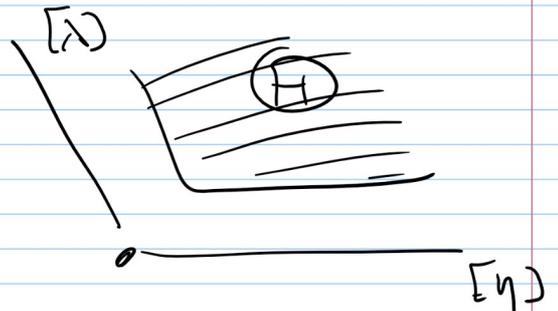
$P_x^n(\log(ghg^{-1}h^{-1})) \in \text{st}^u$
 \uparrow A -norm.

= $[u_2, v_1] +$ terms of at 2 bracket $\in \mathfrak{g}^{[Y]}$

? $[u_2, v_1], u_n$

\uparrow st^u

Hence $[u_2, v_1] \in \text{st}^u$





Proof step 3 By step 2 and linearity

$$\forall u \in \text{Span} \left(\bigcup_{n \geq 1} P_{X^n}(\text{supp } \mu_x^{[n]}) \right)$$

$$\forall v \in \text{Span} \left(\bigcup_{m \geq 1} P_{Y^m}(\text{supp } \mu_y^{[m]}) \right)$$

$$[u, v] \in \text{st}^U$$

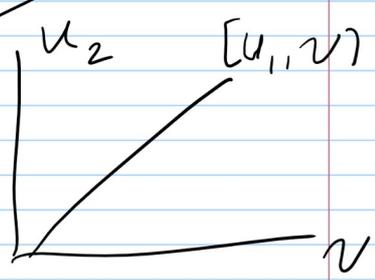
claim if $u_1, u_2 \in$

then $[[u_1, u_2], v] \in \text{st}^U$

Jacobi

$$[[u_1, v], u_2] + [u_1, [u_2, v]]$$

$\in \text{supp } (\mu_x^2)$ by the above.



Next $\forall u, u_2, v_2 \in$

$$[[u, u_2], [u, v_2]] \in \text{st}^U$$

consider

$$\{ (u, v) \in \mathcal{H}^{[1]} \times \mathcal{H}^{[2]} \mid [u, v] \in \text{st}^U \}$$

V-step 3
V-theorem

Theorem Assume \mathcal{H} is simple / \mathbb{R}

• μ as above.

$$\text{If } \exists a \in A. h_\mu(a) > h_{m_x}(a) - \varepsilon(a)$$

$$\varepsilon(a) = d(0, \log \text{Spec}(Ad(a)) \setminus \{0\}) > 0.$$

then μ is invariant and G^0 .

Lemma $V \subsetneq U \subset G_a$ A-norm.

If $\text{supp } \mu_x \subset U$ for a.e. x

then $h_\mu(a, U) - h_{m_x}(a, V) \leq h_{m_x}(a, U) - \varepsilon(a)$

Proof By tower rule. $h_\mu(a, U) = h_\mu(a, V)$

+ assumption $\Rightarrow \mu_x^U \propto \mu_x^V \Rightarrow$

Recall $h_{m_x}(a, V) = -\log \det \text{Ad}(a)|_{\text{Lie } V}$

Lemma $h_\mu(a) = h_\mu(a, G_n) = \sum h_\mu(a, G^{[\lambda]})$

Brain's lecture

$\uparrow \{[\lambda] : \lambda(a) < 1\}$

Consequence of product theorem.



Proof of theorem

Claim $H_x^{[\lambda]} = G^{[\lambda]}$ $\forall \lambda$ with $\lambda(a) < 1$.

because otherwise $h_\mu(a) \leq h_{m_x}(a) - \varepsilon(a)$.

Also $h_\mu(a^{-1}) \geq h_\mu(a) > h_{m_x}(a^{-1}) - \varepsilon(a^{-1})$

So $H_x^{[\lambda]} = G^{[\lambda]}$ $\forall \lambda$ with $\lambda(a) > 1$.

In view of the first theorem, everything reduces to the following.

Lemma Assume \mathfrak{g} simple \mathbb{R} . then.

$[\mathfrak{g}^{[\lambda]}, \mathfrak{g}^{[\eta]}]$

$[\lambda] \neq [\eta] \neq [\lambda^{-1}]$.

$\lambda(a) \neq 1 \neq \eta(a)$.

generators \mathfrak{g} .