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ENEE 765 Lecture 6 (Analysis of deterministic identification)

We need to define some function space.

For $1 \leq p < \infty$, let

$$L_p^m := \left\{ f : [0, \infty) \rightarrow \mathbb{R}^m \mid \begin{array}{l} \int_0^\infty \|f(t)\|_{\mathbb{R}^m}^p dt < \infty \\ \text{measurable} \end{array} \right\}$$

The integral is in the sense of Lebesgue.

$\|\cdot\|_{\mathbb{R}^m}$ denotes ~~and~~ norm on the

Euclidean space \mathbb{R}^m since all such norms are equivalent. [Two norms in \mathbb{R}^m denoted as $\|\cdot\|_a$ and $\|\cdot\|_b$ are equivalent if there exists k_1, k_2, k_3, k_4 all positive such that

$$k_1 \|x\|_a \leq \|x\|_b \leq k_2 \|x\|_a$$

and

$$k_3 \|x\|_b \leq \|x\|_a \leq k_4 \|x\|_b]$$

If the function $f \in L_p^m$, then we define the function space norm

$$\|f\|_p = \left(\int_0^\infty \|f(t)\|_{\mathbb{R}^m}^p dt \right)^{1/p}.$$

$$L_\infty^m := \left\{ f : [0, \infty) \rightarrow \mathbb{R}^m \mid \begin{array}{l} \|f(t)\|_{\mathbb{R}^m} \leq M \text{ for some } M > 0 \\ \text{measurable} \end{array} \right\}$$

essentially bounded $\xrightarrow{\quad}$ except possibly on a set of measure zero

We then define the function space norm to be

$$\|f\|_{\infty} = \text{ess sup}_{t \geq 0} \|f(t)\|$$

on the space L_{∞}^m

It is convenient to drop the superscript m on L_p^m and L_{∞}^m as it will be apparent from the context. Similarly we drop the subscript \mathbb{R}^m from $\|\cdot\|_{\mathbb{R}^m}$ as it will be apparent what is meant from the context.

The aforesaid normed linear spaces are complete [i.e. every Cauchy sequence in such a space converges]. They are referred to as Banach spaces.

When $p=2$, and $m=1$, we recognize an positive definite inner product on L_2 by setting

$$\langle f, g \rangle = \int_0^{\infty} f(t)g(t) dt$$

and

$$\|f\|_2 = \langle f, f \rangle^{1/2}$$

so L_2 is a Hilbert space over the Reals.

We now prove a set of properties of a gradient algorithm from Lecture 5.

(a) Consider $\dot{\phi} = \dot{\theta} = -\gamma e_1 w$ $\gamma > 0$
 $e_1 = \phi^T w$

where $w: \mathbb{R}_+ \rightarrow \mathbb{R}^{2n}$ is assumed to be piecewise continuous. Then

$$e_1 \in L_2 \text{ and } \phi \in L_\infty$$

Proof $\dot{\phi} = -\gamma w w^T \phi \quad (*)$

$$\nabla(\phi) = \frac{1}{2} \phi^T \phi$$

satisfies

$$\dot{\nabla}(\phi) = \frac{d}{dt} \nabla(\phi(t)) \Big|_{(*)}$$

$$\begin{aligned} &= \phi^T \dot{\phi} \\ &= -\gamma \phi^T w w^T \phi \\ &= -\gamma (\phi^T \phi)^2 \\ &\leq 0 \end{aligned}$$

Hence $0 \leq \nabla(\phi(t)) \leq \nabla(\phi(0)) \quad \forall t \geq 0$

$$\Rightarrow \boxed{\phi \in L_\infty}$$

$\nabla(\phi(t))$ is monotone decreasing and is bounded below. Hence $\lim_{t \rightarrow \infty} \nabla(\phi(t))$ exists, is finite, and denoted ~~∇~~ $\nabla_\infty \geq 0$.

But,

$$\begin{aligned}
 \int_0^\infty e_1^2(t) dt &= \int_0^\infty (\phi^T(t) w(t))^2 dt \\
 &= \int_0^\infty -\frac{\dot{\phi}(\phi(t))}{\gamma} dt \\
 &= \frac{V(\phi(0)) - V_\infty}{\gamma} < \infty
 \end{aligned}$$

$$\Rightarrow e_1 \in L_2 \quad \boxed{!}$$

□

(b) As a variant on the gradient algorithm consider the normalized version,

$$\dot{\phi} = \dot{\theta} = -\gamma \frac{e_1 w}{1 + \varepsilon_0 w^T w}; \quad e_1 = \phi^T w$$

with $\varepsilon_0 > 0$, $\gamma > 0$. Assume $w: \mathbb{R} \rightarrow \mathbb{R}^{2n}$
is piecewise continuous. ~~and~~ Thus

$$(i) \quad \frac{e_1}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_2 \cap L_\infty$$

$$(ii) \quad \dot{\phi} \in L_\infty, \quad \dot{\phi} \in L_2 \cap L_\infty$$

$$(iii) \quad \beta = \frac{\phi^T w}{1 + \|w_t\|_\infty} \in L_2 \cap L_\infty$$

(where, $\|w_t\|_\infty = \max_{1 \leq i \leq 2n} |w_i(t)|$

PROOF Let $V(\phi) = \phi^T \phi$

$$\text{Then } \dot{V} = 2\phi^T \dot{\phi} = 2\phi^T \left(-\frac{\gamma e_1 w}{1 + \varepsilon_0 w^T w} \right)$$

$$= \frac{-2 \times \phi^T w w^T \phi}{1 + \varepsilon_0 w^T w} = \frac{-2 \times e_1^2}{1 + \varepsilon_0 w^T w} \leq 0$$

Then $0 \leq V(\phi(t)) \leq V(\phi(0)) \quad \forall t \geq 0$

Hence $V(\phi) \in L_\infty$

$$\|\phi(t)\|_2 = \sqrt{\sum_{i=1}^{2n} (\phi_i(t))^2}$$

$$\leq \sqrt{V(\phi(0))} \quad \forall t \geq 0$$

$$\Rightarrow \boxed{\phi \in L_\infty} \checkmark \text{ (ii)}$$

$$\left\| \frac{e_1(t)}{\sqrt{1 + \varepsilon_0 w^T(t) w(t)}} \right\| = \left\| \frac{\phi^T(t) w(t)}{\sqrt{1 + \varepsilon_0 w^T(t) w(t)}} \right\|$$

$$= \frac{|\phi^T(t) w(t)|}{\sqrt{1 + \varepsilon_0 w^T(t) w(t)}}$$

$$\leq \frac{\|\phi(t)\|_2 \|w(t)\|_2}{\sqrt{1 + \varepsilon_0 w^T(t) w(t)}} \quad (\text{Cauchy-Schwarz})$$

$$\leq \sqrt{V(\phi(0))} \cdot \frac{1}{\sqrt{\varepsilon_0}} \sqrt{\frac{w^T(t) w(t)}{\frac{1}{\varepsilon_0} + w^T(t) w(t)}}$$

$$\leq \frac{1}{\sqrt{\varepsilon_0}} \sqrt{V(\phi(0))}$$

$$\Rightarrow \boxed{\frac{e_1}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_\infty} \checkmark$$

$$|\beta(t)| = \frac{|\phi^T(t) w(t)|}{1 + \max_{1 \leq i \leq 2n} |w_i(t)|}$$

$$\leq \sum_{j=1}^{2n} |\phi_j(t)| \frac{|w_j(t)|}{1 + \max_{1 \leq i \leq 2n} |w_i(t)|}$$

$$\leq \sum_{j=1}^{2n} |\phi_j(t)|$$

$$\leq 2n \|\phi(t)\|_2$$

$$\leq 2n \sqrt{\gamma(\phi(0))}$$

$$\Rightarrow \boxed{\beta \in L_\infty}$$

Thus we have shown that $\gamma, \phi, \frac{e_1}{\sqrt{1 + \varepsilon_0 w^T w}}, \beta$ all belong to L_∞ .

We now need to show the "L₂ part".

$$\dot{\phi} = -\frac{\gamma w e_1}{1 + \varepsilon_0 w^T w}$$

$$= -\frac{\gamma}{\varepsilon_0} \left(\frac{\varepsilon_0 w w^T \phi}{1 + \varepsilon_0 w^T w} \right)$$

aside: for matrix A $\|Ax\|_2 = \sqrt{x^T A^T A x}$

$$\leq \sqrt{\lambda_{\max}(A^T A)} \sqrt{x^T x}$$

$$= \sqrt{\lambda_{\max}(A^T A)} \|x\|_2$$

Using the above (Aside),

let $A = WW^T$.

$$\begin{aligned}\lambda_{\max}(A^T A) &= \lambda_{\max}(WW^T W W^T) \\ &= \lambda_{\max}((W^T W) W W^T) \\ &= (W^T W)^2\end{aligned}$$

Hence

$$\begin{aligned}\|WW^T \phi\|_2 &\leq \sqrt{\lambda_{\max}((WW^T)^T (WW^T))} \|\phi\|_2 \\ &= \sqrt{(W^T W)^2} \|\phi\|_2 \\ &= W^T W \|\phi\|_2 \quad (\text{all evaluated at } t)\end{aligned}$$

$$\begin{aligned}\Rightarrow \|\dot{\phi}(t)\|_2 &\leq \frac{\gamma}{\varepsilon_0} \frac{(\varepsilon_0 W^T(t) W(t))}{(1 + \varepsilon_0 W^T(t) W(t))} \|\phi(t)\|_2 \\ &\leq \frac{\gamma}{\varepsilon_0} \|\phi(t)\|_2 \\ &\leq \frac{\gamma}{\varepsilon_0} \sqrt{V(\phi(0))}\end{aligned}$$

$$\Rightarrow \boxed{\dot{\phi} \in L_\infty}$$

$$\begin{aligned}\int_0^\infty \left[\frac{e_1(t)}{\sqrt{1 + \varepsilon_0 W^T(t) W(t)}} \right]^2 dt &= \int_0^\infty \frac{e_1^2(t)}{1 + \varepsilon_0 W^T(t) W(t)} dt \\ &= \int_0^\infty \frac{\dot{V}(\phi(t))}{2\gamma} dt = \frac{V(\phi(0)) - V_\infty}{2\gamma} \quad \text{exist and is finite}\end{aligned}$$

Thus $\frac{e_i}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_2$

We have thus far shown ✓

$$\frac{e_i}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_2 \cap L_\infty \quad (\text{i})$$

$$\beta(t) = \frac{\phi^T(t) w(t)}{1 + \max_{1 \leq i \leq n} |w_i(t)|}$$

$$= \frac{\phi^T(t) w(t)}{\sqrt{1 + \varepsilon_0 w^T(t) w(t)}} \cdot \frac{\sqrt{1 + \varepsilon_0 w^T(t) w(t)}}{1 + \max_{1 \leq i \leq n} |w_i(t)|}$$

This belongs to L_2
(see above)

if we show this belongs to
 L_∞ then we can conclude

$$\beta \in L_2$$

Recall, in finite dimensions, say in \mathbb{R}^k

$$\|x\|_\infty = \max_{1 \leq i \leq k} \|x_i\| \leq \|x\|_2 \leq \sqrt{k} \|x\|_\infty$$

Hence

$$\frac{\sqrt{1 + \varepsilon_0 \|w(t)\|_2^2}}{1 + \max_{1 \leq i \leq 2n} |w_i(t)|} \leq \frac{\sqrt{1 + \varepsilon_0 \|w(t)\|_2^2}}{1 + \frac{1}{\sqrt{2n}} \|w(t)\|_2}$$

The right hand side of the above inequality has the form, $\sqrt{f(y)}$, where with $y \geq 0$

$$\begin{aligned} f(y) &= \frac{1 + \varepsilon_0 y^2}{\left(1 + \frac{1}{\sqrt{2n}} y\right)^2} \\ &\approx \frac{1 + \varepsilon_0 y^2}{1 + \frac{1}{2n} y^2 + \sqrt{\frac{2}{n}} y} \end{aligned}$$

$$\rightarrow 2n\varepsilon_0 \text{ as } y \rightarrow \infty.$$

Pick a $\delta > 0$. Then $\exists Y > 0$ such that, $\forall y > Y$, $f(y) < 2n\varepsilon_0 + \delta$. On the other hand, on the closed and bounded set $[0, Y]$, $f(y)$ has a maximum f_{\max} due to continuity of $f(y)$ - Weierstrass Theorem.

Hence $f(y) < \max(f_{\max}, 2n\varepsilon_0 + \delta)$

Thus $\sqrt{f} \in L_\infty$. Thus we have shown $\beta \in L_2$
we have now shown $\beta \in L_2 \cap L_\infty$ (iii) ✓

$$\begin{aligned}
 \|\dot{\phi}(t)\|_2^2 &= \frac{\gamma^2 e_1^2(t) w^T(t) w(t)}{(1 + \varepsilon_0 w^T(t) w(t))^2} \\
 &= \frac{\gamma^2}{\varepsilon_0} \frac{e_1^2(t)}{(1 + \varepsilon_0 w^T(t) w(t))} \frac{w^T(t) w(t)}{(1 + \varepsilon_0 w^T(t) w(t))} \\
 &\leq \frac{\gamma^2}{\varepsilon_0} \frac{e_1^2(t)}{(1 + \varepsilon_0 w^T(t) w(t))}
 \end{aligned}$$

We have already shown $\frac{e_1}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_2$

Hence $\dot{\phi} \in L_2$ which then $\Rightarrow \boxed{\dot{\phi} \in L_2 \cap L_\infty \text{ (ii)}}$ □

(c) Suppose $\dot{\phi} = \dot{\theta} = -\gamma e_1 w$

$$e_1 = \phi^T w + \varepsilon \quad \text{(accounting for observer)} \\ \text{(gap/error)}$$

where $\varepsilon(t) \rightarrow 0$ exponentially as $t \rightarrow \infty$

Then $e_1 \in L_2$, $\phi \in L_\infty$

as

PROOF: Define $V(\phi) \leq \phi^T \phi + \frac{\gamma}{2} \int \varepsilon^2(\sigma) d\sigma$

$$\begin{aligned}
 \dot{V} &= 2\phi^T \dot{\phi} - \frac{\gamma}{2} \varepsilon^2(t) \\
 &= 2\phi^T (-\gamma (\phi^T w + \varepsilon) w) - \frac{\gamma}{2} \varepsilon^2 \\
 &= -2\gamma (\phi^T w)^2 - 2\gamma (\phi^T w) \varepsilon - \frac{\gamma}{2} \varepsilon^2 \\
 &= -2\gamma \left(\phi^T w + \frac{\varepsilon}{2} \right)^2 \leq 0
 \end{aligned}$$

$0 \leq V(\phi(t))$ and $V(\phi(t))$ is monotone decreasing with increasing $t \Rightarrow \lim_{t \rightarrow \infty} V(\phi(t))$ exists, is finite $= V_\infty \geq 0$

$$\begin{aligned}\|\phi(t)\|_2 &\leq \sqrt{V(\phi(0))} \\ &= \left(\|\phi(0)\|_2^2 + \int_0^\infty \varepsilon^2(s) ds \right)^{1/2}\end{aligned}$$

$$\Rightarrow \boxed{\phi \in L_\infty}$$

$$\begin{aligned}&\int_0^\infty \left(\phi^T(t) w(t) + \frac{\varepsilon}{2}(t) \right)^2 dt \\ &\quad = \int_0^\infty -\frac{\dot{V}(\phi(t))}{2} dt \\ &\quad = \frac{V(\phi(0)) - V_\infty}{2} \\ &\quad < \infty\end{aligned}$$

$$\text{Thus } \phi^T w + \frac{\varepsilon}{2} \in L_2$$

On the other hand $\frac{\varepsilon}{2} \in L_2$ (since by hypothesis $\varepsilon(t)$ decays exponentially to 0 as $t \rightarrow \infty$).

$$\begin{aligned}\text{Hence } e_1 &= \left(\phi^T w + \frac{\varepsilon}{2} \right) + \frac{\varepsilon}{2} \in L_2 \\ (\text{recall } L_2 &\text{ is a vector space}).\end{aligned}$$

□

Some recapitulation

In proving the properties listed under (a) (b) (c) above, we have used

- (1) $w(\cdot)$ is piecewise continuous \Rightarrow differential equation $\dot{\phi} = -\lambda w w^T \phi$ has well-defined (series) solution.
- (2) A monotone decreasing function $g: [0, \infty) \rightarrow \mathbb{R}$ which satisfies a lower bound $g(t) \geq c$ for c a constant, has a limit
- (3) Cauchy-Schwarz inequality

- (4) for a symmetric matrix Q of size $k \times k$,

$$\lambda_{\min}(Q) x^T x \leq x^T Q x \leq \lambda_{\max}(Q) x^T x$$

$\forall x \in \mathbb{R}^n$.

[the ratio $\frac{x^T Q x}{x^T x}$ is called

Rayleigh quotient]

- (5) If $f \in L_2$, $g \in L_\infty$ with $\|g\| = M$ then

$$f \cdot g \in L_2$$

$$\text{since } \int_0^\infty ((f \cdot g)(t))^2 dt \leq \int_0^\infty f^2(t) \cdot M^2 dt \leq M^2 \int_0^\infty f^2(t) dt$$

In the proof of parts of properties (b), we follows the sequence of proofs

$$\begin{array}{c}
 \text{(ii)} \quad \phi \in L_\infty \\
 \text{(i)} \quad \frac{e_1}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_\infty \quad \frac{e_1}{\sqrt{1 + \varepsilon_0 w^T w}} \in L_2 \\
 \text{(iii)} \quad \beta \in L_\infty \\
 \text{(ii)} \quad \dot{\phi} \in L_\infty \quad \phi \in L_2
 \end{array}$$

$\beta \in L_2$

In what follows we seek state a stability theorem for our deterministic identifier(s) under additional hypotheses. We need a technical result of wide applicability in control theory known as Barbalat's Lemma (L. Barbalat, "Systèmes d'équations différentielles d'oscillations non linéaires", Rev. Math. Pures Appl. 4 267–270 (1959)), and a corollary to it.

Borsbalt's Lemma

A good lemma is worth a thousand theorems

— Doron Zeilberger
Rutgers University

Suppose $f : \mathbb{R} \rightarrow \mathbb{R}$ is uniformly continuous on $[0, \infty)$. Suppose that

$$\lim_{t \rightarrow \infty} \int_0^t f(\sigma) d\sigma$$

exists and is finite. Then

$$f(t) \rightarrow 0 \quad \text{as } t \rightarrow \infty.$$

Proof: We prove by contradiction.

Suppose there is a constant $k_1 > 0$ s.t. for every $T > 0$, we can find $T_1 \geq T$ such that $|f(T_1)| \geq k_1$. Since f is uniformly continuous, there is a positive constant k_2 s.t.

$$|f(t + \varepsilon) - f(t)| < \frac{k_1}{2} \quad \forall t \geq 0$$

and $\underline{\forall} 0 \leq \varepsilon \leq k_2$.

Hence,

$$\begin{aligned} |f(t)| &= |f(t) - f(T_1) + f(T_1)| \\ &\geq |f(T_1)| - |f(t) - f(T_1)| \\ &> k_1 - \frac{k_1}{2} = \frac{k_1}{2} \end{aligned}$$

$$\forall t \in [T_1, T_1 + k_2]$$

Therefore,

$$\left| \int_{T_1}^{T_1+k_2} f(t) dt \right| = \int_{T_1}^{T_1+k_2} |f(t)| dt \\ > \frac{1}{2} k_1 k_2$$

where the equality holds since $f(t)$ has same sign $\forall t \in [T_1, T_1+k_2]$.

Thus $\int_0^t f(\sigma) d\sigma$ cannot converge to

finite limit as $t \rightarrow \infty$, a contradiction. This completes the proof \square

Corollary

Let $g, \dot{g} \in L_\infty$ and $g \in L_p$ for $1 \leq p < \infty$. Then $g(t) \rightarrow 0$ as $t \rightarrow \infty$

Proof Let $f(t) = |g(t)|^p$. Then g, \dot{g} bounded $\Rightarrow f$ is uniformly continuous. $g \in L_p \Rightarrow \lim_{t \rightarrow \infty} \int_0^t f(\sigma) d\sigma$ exists and is finite and $= (\|g\|_p)^p$

By Baire-Banach's Lemma $f(t) \rightarrow 0$ as $t \rightarrow \infty$
Equivalently $g(t) \rightarrow 0$ as $t \rightarrow \infty$ \square

Convergence / Stability Theorem for Identifier,

Consider the deterministic identification problem for a linear, time-invariant, finite-dimensional, strictly proper plant subject to a reference input signal $r(t)$ which is piecewise continuous and bounded.

Suppose that the plant is stable (all poles in the open left half-plane \mathbb{C}^-) or placed in a known linear feedback loop such that r and y_p are bounded. Then, for the gradient algorithm with or without normalization, or the least squares algorithm with normalization and re-setting, the following hold:

- (i) $e_1 \in L_2 \cap L_\infty$
- (ii) $e_1 \rightarrow 0$ as $t \rightarrow \infty$
- (iii) $\dot{\phi}, \phi \in L_\infty$
- (iv) $\dot{\phi} \in L_2 \cap L_\infty$
and
 $\dot{\phi} \rightarrow 0$ as $t \rightarrow \infty$.

Proof In properties (a) (b)(c) above we have shown for the gradient and normalized gradient algorithms,
 $e_1 \in L_2$ $\phi \in L_\infty$

r and y_p bounded \Rightarrow w and \dot{w} bounded

$$\phi, \dot{\phi} \in L_\infty \Rightarrow e_1 = \phi^T w \text{ and}$$

$$\dot{e}_1 = \dot{\phi}^T w + \phi^T \ddot{w} \in L_\infty$$

$$e_1 \in L_2 \text{ and } \dot{e}_1, e_1 \in L_\infty$$

$$\Rightarrow e_1(t) \rightarrow 0 \text{ as } t \rightarrow \infty$$

(by Barbalat's Lemma and its corollary above).

Similarly

$$\phi \in L_2 \text{ and } \phi, \dot{\phi} \in L_\infty$$

$$\Rightarrow \phi(t) \rightarrow 0 \text{ as } t \rightarrow \infty$$

(Here we need to appeal to Barbalat's lemma with functions taking values in 2h dimensional vector spaces - an easy generalization) \square