

# A Resolvent Framework for Global and Local Nonlinear Semigroups

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## Symbols

$X$	Banach space $(X,    \cdot   )$
$w(x, x_t, \theta) : X \times X \times \mathbb{R} \rightarrow \mathbb{R}$	Weighting kernel at $x_t$ , parameterized by $\theta \in \mathbb{R}, x_t \in X$

## 1 Definitions

**Definition 1.** *The process- $t$  matrix, read “process until  $t$  matrix” or more simply “process until  $t$ ”, is*

$$X_t = \begin{bmatrix} x_t \\ x_{t-h} \\ x_{t-2h} \\ \vdots \end{bmatrix}$$

where rows belong to  $X$ .

**Definition 2.** The  $t$ -weighting matrix of the process- $i$  is given by

$$W(X_i, x_t, \theta) = \begin{bmatrix} w(x_j, x_t, \theta) & 0 & 0 & \cdots \\ 0 & w(x_{j-h}, x_t, \theta) & 0 & \cdots \\ 0 & 0 & w(x_{j-2h}, x_t, \theta) & \cdots \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

$W_t^\theta X_i$  is short hand for the the product  $W(X_i, x_t, \theta)X_i$

$$W_t^\theta X_i := W(X_i, x_t, \theta) \cdot X_i = \begin{bmatrix} w(x_i, x_t, \theta) \cdot x_i \\ w(x_{i-h}, x_t, \theta) \cdot x_{i-h} \\ w(x_{i-2h}, x_t, \theta) \cdot x_{i-2h} \\ \vdots \end{bmatrix}$$

which is the  $t$ -weighting of the process- $i$ .

## 2 Introduction

### 2.1 $C_t$ as a globally linear map

Consider the map

$$C : x_t \mapsto x_{t+h}$$

for  $x_i \in X_t \subset X$  and

$$x_{t+h} = Cx_t$$

$C$  may be state dependent,  $C(x(t))$ , non-autonomous,  $C(t)$ , or both,  $C(t, x(t))$ . We denote the possibility of any such case as  $C_t$ .

If  $C_t$  is globally linear then

$$X_{t+h} = X_t C_t$$

which is solved via,

$$C_t = X_t^{-1} X_{t+h}$$

In an applied time series setting this solution for  $C_t$  is the *auto-regressive* or *AR* model for the process  $X_t$ .

### 2.2 $C_t^\theta$ as a locally linear map

If  $C$  is not globally linear than the  $t$ -weighting can be introduced such that

$$W_t^\theta X_{t+h} = W_t^\theta X_t C_t^\theta$$

where the weighting kernel,  $w$ , of  $W_t^\theta$  is parameterized by  $\theta$ . The exact local weighting at  $x_t$  is achieved as,

$$\lim_{\theta \rightarrow \infty} w(x_i, x_t, \theta) = \delta(x_i - x_t)$$

for all  $x_i \in X_t$ . Typically the kernel is chosen so  $w \sim e^{-\theta}$ . The solution for  $C_t^\theta$  is

$$\begin{aligned} C_t^\theta &= (W_t^\theta X_t)^{-1} W_t^\theta X_{t+h} \\ &= X_t^{-1} (W_t^\theta)^{-1} W_t^\theta X_{t+h} \end{aligned}$$

Taking the limit,  $W_t^\theta$  reduces to,

$$\lim_{\theta \rightarrow \infty} W_t^\theta = \begin{cases} \delta(x_i - x_t) = 1 & \text{if } i = j \wedge x_i = x_t \\ 0 & \text{if } i \neq j \vee x_i \neq x_t \end{cases}$$

which is a diagonal matrix whose only non-zero entries are 1's corresponding to states arbitrarily close to  $x_t$ . If  $X_t$  never returns to states arbitrarily close to  $x_t$  then  $W_t^\theta$  reduces to the rank-1 matrix with a single non-zero entry,

$$(W_t^\theta)_{1,1} = 1$$

If  $X_t$  is periodic at frequency  $k$  the set of all indices  $i$  on the diagonal where  $(W_t^\theta)_{i=j} = 1$ , is

$$\{W_t^\theta\}_1 = \{n \in \mathbb{N} : i = 1 + nk\}$$

If  $X_t$  is ergodic than the rank depends on the nature and frequency of close returns to neighbourhoods containing  $x_t$ , e.g.

$$U = \{x \in X, \delta = a \in \mathbb{R} : w(x, x_t, \theta) < \delta\}$$

could be used to define “close”.

### 3 One Parameter Semigroups

Placing  $C_t^\theta$  in the theory of semigroups can help us understand the relationship between locality,  $\theta$ , and the potential higher order terms for some generating process of the elements of  $X_t$ . The study of one parameter semigroups provides useful tools that connect the operators which generate the semigroup to the semigroups themselves, by way of *generating theorems*. We start by generalizing  $C_t$  as a semigroup.

#### 3.1 $C_t$ as a one parameter semigroup

We define  $C_t$

$$C(t) : t \rightarrow \mathbb{G} \qquad t \in \mathbb{R}_{\geq 0}$$

where we take the group  $\mathbb{G}$  as  $\{X_t\}_{t \geq 0}$

$$C(t) : t \rightarrow X_{t+h} \qquad X_{t+h} \in X$$

with  $(X, \|\cdot\|)$  being Banach.

$\{C(t)\}_{t \geq 0}$  forms a one parameter semigroup if the the following requirements are met:

Associativity:

$$\forall t, s \geq 0 : C(t + s) = C(t)C(s) \quad (1)$$

Identity:

$$C(0) = I \quad (2)$$

### 3.2 $C_0$ -semigroup

Continuity for one parameter semigroups are often defined by the following properties:

Strongly Continuous:

For the *infinitesimal generator*,  $A$ , defined on the domain  $D(A)$

$$Ax = \lim_{t \downarrow 0} \frac{1}{t} (C(t) - I)x \quad (3)$$

the limit must exist.

The strongly continuous one parameter semigroups is denoted as the  $C_0$ -semigroup. This semigroup provides a generalization of the exponential in the case where it is also uniformly continuous;

Uniform continuity:

$$\lim_{t \downarrow 0} \|C(t)x(0) - x(0)\| \rightarrow 0 \quad \forall x(0) \in X \quad (4)$$

The linear operator  $A$  is the infinitesimal generator of a uniformly continuous semigroup if and only if  $A$  is a bounded linear operator. In this case the infinitesimal generator  $A$  of the semigroup  $C(t)$  must satisfy

$$C(t) = e^{At} := \sum_0^{\infty} \frac{A^k}{k!} t^k$$

However if  $C(t)$  is strongly but not uniformly continuous then  $A$  is not bounded and  $e^{At}$  need not converge.

The  $C_0$ -semigroup provides a good, however insufficient, basis for understanding the relationships of locality and the higher order terms of the generator for  $X_t$ .

### 3.3 $C_0^n$ -semigroup

$C_0$ -semigroup to be a special case of a more general strongly continuous one parameter semigroup, the nonlinear  $C_0^n$ -semigroup, whose infinitesimal generator expresses higher order effects in the limit of ever smaller finite differences.

We both guide and justify the definition such a semigroup on the basis that the exponential of the infinitesimal generator yields a well known generalization of the Taylor series that uses the infinitesimal limit of finite differences. This is the Hille series.

The upshot is that the  $C_0^n$ -semigroup and its infinitesimal generator can carry intuition regarding continuity and nonlinearity over to pragmatic operator theory approaches, wherein we are better suited to investigate locality.

### 3.3.1 The Hille Operator for nonlinear semigroups

We can relate the forward differences to the map of  $C$  where,

$$\Delta_h^1 x(t) = x(t+h) - x(t) = C_h^1 x(t) - x(t)$$

.

For higher order differences the difference operator series is

$$\begin{aligned} \Delta_h^n x(t) &= \sum_{k=0}^n (-1)^{n-k} \binom{n}{k} x(t+kh) \\ &= \sum_{k=0}^n (-1)^{n-k} \binom{n}{k} C_h^k x(t) \end{aligned} \tag{5}$$

If we define the  $n$ -order generator,

$$A_h^n = \frac{\Delta_h^n}{h^n}$$

The  $n$ -order generators of the semigroup,  $A_h^n$ , describe higher-order terms of the expansion when viewed as finite-difference corrections to the first-order dynamics,  $A_h^1$ , capturing more detailed behavior of the system.

A semigroup can be defined over the sum of the  $n$ -order generators,

$$\begin{aligned} C_h(t) &= e^{A_h t} := \sum_{n=0}^{\infty} \frac{t^n}{n!} A_h^n \\ &:= \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{\Delta_h^n}{h^n} \end{aligned} \tag{6}$$

In the limit of difference size  $h$  we get

$$\begin{aligned} \lim_{h \rightarrow 0^+} C_h(t) &= \lim_{h \rightarrow 0^+} \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{\Delta_h^n}{h^n} \\ H(t) &= \lim_{h \rightarrow 0^+} \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{\Delta_h^n}{h^n} \end{aligned}$$

where we call  $H(t)$  the *Hille operator*. On application to  $x(t)$ , this operator yields the Hille series expansion

$$H(t)x(t) = \lim_{h \rightarrow 0^+} \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{\Delta_h^n x(t)}{h^n}$$

Thus, the Hille Operator provides a way to generalize the semigroup dynamics to include higher-order effects via the Hille Series, defining a strongly and uniformly continuous one parameter nonlinear semigroup. We denote the order of the semigroup as  $C_0^n$ -semigroup. Where  $n$  is non finite we have the strongly nonlinear  $C_0^\infty$ -semigroup.

We can reintroduce (5) so,

$$\begin{aligned} H(t) &= \lim_{h \rightarrow 0^+} \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{\Delta_h^n}{h^n} \\ &= \lim_{h \rightarrow 0^+} \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{\sum_{k=0}^n (-1)^{n-k} \binom{n}{k} C_h^k}{h^n} \end{aligned}$$

### 3.4 The resolvent of $C_0^n$

The resolvent

$$R(\lambda, H(t)) = (\lambda I - H(t))^{-1}$$

is important because it tells us whether the Hille operator  $H(t)$  has an inverse for certain values of  $\lambda$ . This can help define the spectrum and growth behavior of the  $C_0^n$  semigroup.

We are interested in the resolvent set that provides a well defined inverse,

$$\rho(H) = \left\{ n \in \mathbb{Z}, \lambda \in \mathbb{C} : \lim_{h \rightarrow 0^+} (\lambda I - H(t))^{-1} \right\}$$

We have

$$\begin{aligned} R(\lambda, H(t)) &= \lim_{h \rightarrow 0^+} (\lambda I - H(t))^{-1} \\ &= \lim_{h \rightarrow 0^+} \left( \lambda I - \sum_n \frac{t^n}{n!} \frac{A_h^n}{h^n} \right)^{-1} \end{aligned}$$

so,

$$0 \neq \lim_{h \rightarrow 0^+} |\lambda I - H(t)|$$

We want  $\lambda \in \rho(H(t))$  where

$$\begin{aligned} \rho(H(t)) &= \mathbb{C} / \sigma(H(t)) \\ &= \mathbb{C} / \sigma \left( \lim_{h \rightarrow 0^+} \sum_n \frac{t^n}{n!} \frac{A_h^n}{h^n} \right) \end{aligned}$$

There are a number of options to resolve  $\sigma(H(t))$  in non-empirical settings. These include, among others, methods from functional calculus, operator, theory, and perturbation theory.

## 4 An Empirical Hille Operator

In an empirical setting the coefficients,  $a_n$ , of the operator,

$$H(t) = \lim_{h \rightarrow 0^+} \sum_n a_n \frac{\Delta_h^n}{h^n}$$

may not ensure convergence. In such cases we have to consider how to treat the distribution of coefficients,

$$P(a_n) \sim f(n)$$

assuming it is some function of the order  $n$ . In this case the only required constraint is normalization

$$\sum_n P(a_n) = 1$$

Beyond this many natural systems may or may not exhibit further constraints including bounded higher order moments. How should we, *a priori*, evaluate methods to estimate  $P(a_n)$

If we seek the least biased estimate of the dynamics described by  $H(t)$  then we should adopt a maximum entropy approach. Following such a solution we may assess the spectra of the ensemble,  $\sigma(< H(t)x(t) >)$ , where

$$< H(t)x(t) > = \sum_n P(a_n) a_n \frac{\Delta_h^n x(t)}{h^n}$$

#### 4.1 Maximum Entropy Estimates of $H(t)$

Given a set of trajectory data,  $\{X_t\}_t \geq 0$ , we propose the following application of the empirical Hille operator,

$$\begin{aligned} H(t)x(t) &= \sum_n P(a_n) a_n \frac{\Delta_h^n x(t)}{h^n} \\ &= \sum_n P(a_n) a_n \frac{\sum_{k=0}^n (-1)^{n-k} \binom{n}{k} C_h^k(t)}{h^n} \\ &= \sum_n P(a_n) a_n \frac{\sum_{k=0}^n (-1)^{n-k} \binom{n}{k} X_t^{-1} X_{t+kh}}{h^n} \\ &= \sum_n X_t^{-1} P(a_n) a_n \frac{\sum_{k=0}^n (-1)^{n-k} \binom{n}{k} X_{t+kh}}{h^n} \end{aligned}$$

The entropy of  $P(a_n)$  is given by

$$S(P) = - \sum_n P(a_n) \log P(a_n)$$

where,

$$\sum_n P(a_n) = 1$$

We seek the solution to the Lagrangian,

$$\mathcal{L} = - \sum_n P(a_n) \log P(a_n) + \lambda \left( \sum_n P(a_n) - 1 \right)$$