

# Detecting systematic anomalies affecting systems when inputs are stationary time series

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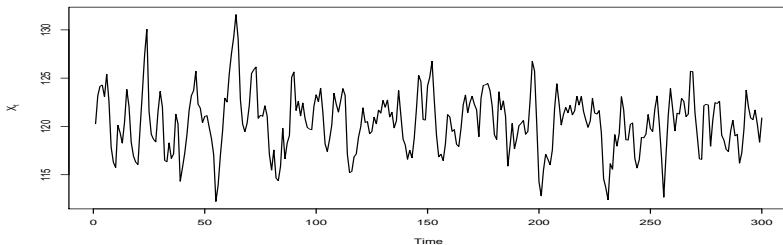
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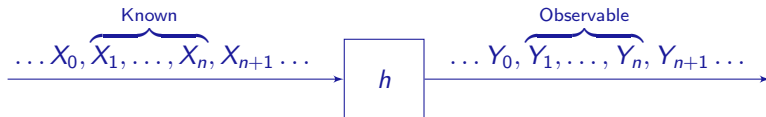
March 10, 2021

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Life as usual: ups and downs...

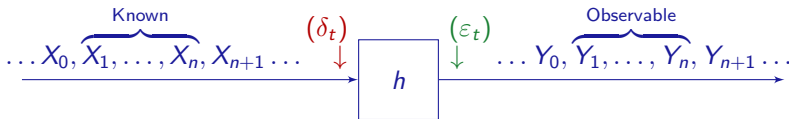


...or is it?



- Inputs  $(X_t)_{t \in \mathbb{Z}}$
- Transfer function  $h$ 
  - ↗ Ex 1: Automatic voltage regulator
  - ↘ Ex 2: Insurer processing claims
- Outputs  $(Y_t)_{t \in \mathbb{Z}}$
- Observable pairs  $(X_1, Y_1), \dots, (X_n, Y_n)$

Think of  $(X_t)_{t \in \mathbb{Z}}$  as a stationary and causal time series



- Input risks  $(\delta_t)_{t \in \mathbb{Z}}$

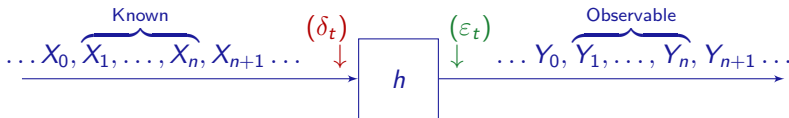
↗ measurement errors

Think of risks as → miscalculations

- Output risks  $(\varepsilon_t)_{t \in \mathbb{Z}}$

↘ oversights

- $$Y_t = \begin{cases} h(X_t + \delta_t) & \text{when only inputs are directly affected} \\ h(X_t) + \varepsilon_t & \text{when only outputs are affected} \\ h(X_t + \delta_t) + \varepsilon_t & \text{when inputs \& outputs are affected} \end{cases}$$



If  $h$  were known, we could use, e.g.,

$$\frac{1}{n} \sum_{i=1}^n (h(X_i) - Y_i)^2 \quad \begin{cases} = 0 & \text{when the system is risk free} \\ > 0 & \text{when the system is risk affected} \end{cases}$$

But we only know that  $h \in \mathcal{H}$  (model uncertainty)

## Our philosophy

We check (for risks) only those systems that **were in reasonable order** when newly installed

What does “**be in reasonable order**” mean?

Whose definition to use?

- **Manufacturer's definition?** Maybe, but likely only indirectly
- **Our definition?** Yes, because it is aligned with our goals

...and it is on the next slide

**Definition.** The risk-free outputs  $Y_t = h(X_t)$  are in reasonable order if

$$B_n^0 := \frac{1}{n^{1/2}} \sum_{i=2}^n |h(X_{i:n}) - h(X_{i-1:n})| = O_{\mathbb{P}}(1)$$

where  $X_{1:n} \leq \dots \leq X_{n:n}$  are the ordered inputs  $X_1, \dots, X_n$

**Example** (to work out intuition). If  $h \in \text{Lipschitz}$ , then the risk-free outputs are in reasonable order because

$$\begin{aligned} B_n^0 &\leq \frac{\|h\|_{\text{Lip}}}{n^{1/2}} \sum_{i=2}^n |X_{i:n} - X_{i-1:n}| = \|h\|_{\text{Lip}} \frac{X_{n:n} - X_{1:n}}{n^{1/2}} \\ &= \|h\|_{\text{Lip}} \frac{\text{Range}(X_1, X_2, \dots, X_n)}{n^{1/2}} = O_{\mathbb{P}}(1) \end{aligned}$$

Suppose that a brand new system was in reasonable order

- If it is still **risk free**, then  $I_n \rightarrow_{\mathbb{P}} \text{somewhere} \neq 0.5$
- If it gets **risk affected**, then  $I_n \rightarrow_{\mathbb{P}} 0.5$

What is this magical  $I_n$ ?

The system-monitoring index  
(statistic)

$$I_n = \frac{\sum_{i=2}^n (Y_{i,n} - Y_{i-1,n})_+}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|}$$

where  $Y_{1,n}, \dots, Y_{n,n}$  are the  
**concomitants** of  $X_1, \dots, X_n$

### Example

$(X_i, Y_i)$	$(X_{i,n}, Y_{i,n})$
(5, 3)	( <b>1</b> , <b>9</b> )
(1, 9)	( <b>2</b> , <b>6</b> )
(4, 2)	( <b>4</b> , <b>2</b> )
(2, 6)	( <b>5</b> , <b>3</b> )



A mathematical insight into the meaning of  $I_n$ 

$$I_n = \frac{\sum_{i=2}^n (Y_{i,n} - Y_{i-1,n})_+}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|} = \frac{1}{2} \left( 1 + \frac{Y_{n,n} - Y_{1,n}}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|} \right)$$

because  $x_+ = (|x| + x)/2$

$$= \frac{1}{2} \left( 1 + \frac{\text{Pseudo Range}(Y_1, Y_2, \dots, Y_n)}{\text{Total Variation}(Y_1, Y_2, \dots, Y_n)} \right)$$

if there are no risks  
and  $n$  is large

$$\approx \frac{1}{2} \left( 1 + \frac{\int h'(x) dx}{\int |h'(x)| dx} \right)$$

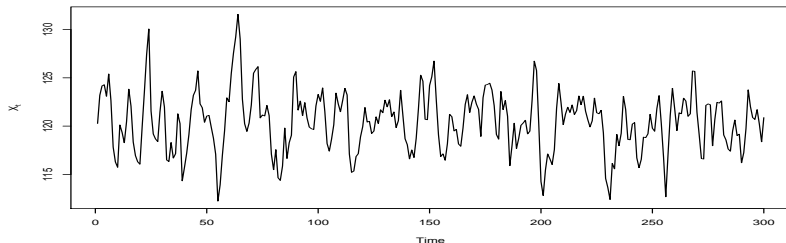
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**Example.** Let risk free  $(X_t)$  be ARMA(1, 1) and follow

$$(X_t - 120) = 0.6(X_{t-1} - 120) + \eta_t + 0.4\eta_{t-1}$$

with iid Gaussian innovations  $\eta_t \sim \mathcal{N}(0, \sigma_\eta^2)$  where  $\sigma_\eta^2$  is such that

$$X_t \sim \mathcal{N}(120, 9)$$



Automatic voltage regulators (AVR's) often keep voltage between

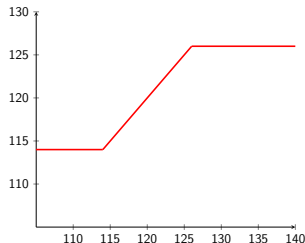
$120 \pm 6$  volts ( $\pm 5\%$  of the nominal voltage)

The transfer function

$$h(x) = \begin{cases} x_{\min} & \text{when } x < x_{\min} \\ x & \text{when } x_{\min} \leq x \leq x_{\max} \\ x_{\max} & \text{when } x > x_{\max} \end{cases}$$

with

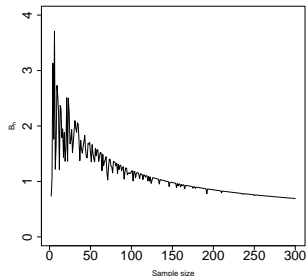
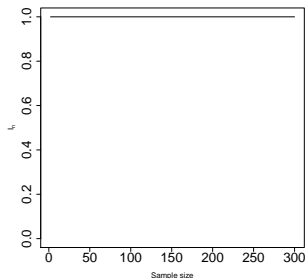
$$x_{\min} = 114 \quad \& \quad x_{\max} = 126$$



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**Note.** Insurance layers often have similar transfer functions: deductible  $x_{\min}$ , policy limit  $x_{\max}$ , etc

When the risk-free AVR is in reasonable order, we see



$$I_n^0 = \frac{\sum_{i=2}^n (h(X_{i:n}) - h(X_{i-1:n}))_+}{\sum_{i=2}^n |h(X_{i:n}) - h(X_{i-1:n})|}$$
$$= 1 \neq 0.5 \quad (\text{risk free})$$

$$B_n^0 = \frac{1}{n^{1/2}} \sum_{i=2}^n |h(X_{i:n}) - h(X_{i-1:n})|$$
$$= O_{\mathbb{P}}(1) \quad (\text{in reasonable order})$$

Illustrative risk specifications: let  $(\delta_t)$  and  $(\epsilon_t)$  be

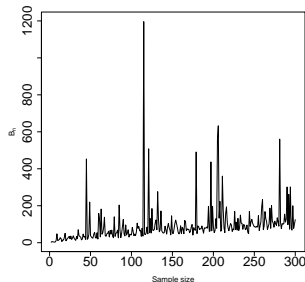
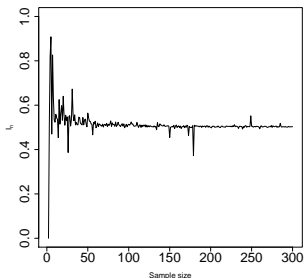
- independent of  $(X_t)$
- independent of each other
- iid  $\text{Lomax}(\alpha, 1)$  and thus have the means

$$\mathbb{E}(\delta_t) = \mathbb{E}(\epsilon_t) = \frac{1}{\alpha - 1}$$

Examples:  $\alpha = 1.2 \implies \mathbb{E}(\delta_t) = \mathbb{E}(\epsilon_t) = 5$

$$\alpha = 11 \implies \mathbb{E}(\delta_t) = \mathbb{E}(\epsilon_t) = 0.1$$

$$\mathbb{E}(\delta_t) = \mathbb{E}(\epsilon_t) = 5 \quad (\text{fast converging } I_n \text{ and } B_n)$$



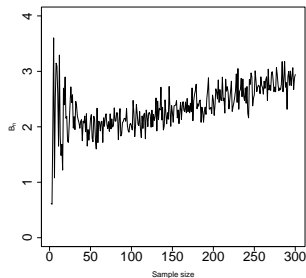
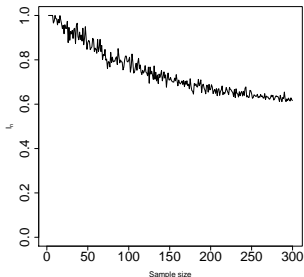
$$I_n = \frac{\sum_{i=2}^n (Y_{i,n} - Y_{i-1,n})_+}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|}$$

→  $\mathbb{P}$  0.5 (risk affected)

$$B_n = \frac{1}{n^{1/2}} \sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|$$

→  $\mathbb{P}$   $\infty$  (out of reasonable order)

$$\mathbb{E}(\delta_t) = \mathbb{E}(\epsilon_t) = 0.1 \quad (\text{slowly converging } I_n \text{ and } B_n)$$



$$I_n = \frac{\sum_{i=2}^n (Y_{i,n} - Y_{i-1,n})_+}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|}$$

$\rightarrow \mathbb{P} \ 0.5$  (risk affected)

$$B_n = \frac{1}{n^{1/2}} \sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|$$

$\rightarrow \mathbb{P} \ \infty$  (out of reasonable order)



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The answer depends on the validity of

$$I_n = \frac{\sum_{i=2}^n (Y_{i,n} - Y_{i-1,n})_+}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|} \rightarrow_{\mathbb{P}} I_{\infty} \begin{cases} = 0.5 & \text{when risk affected} \\ \neq 0.5 & \text{when risk free} \end{cases}$$

assuming

$$B_n^0 = \frac{1}{n^{1/2}} \sum_{i=2}^n |h(X_{i,n}) - h(X_{i-1,n})| = O_{\mathbb{P}}(1)$$

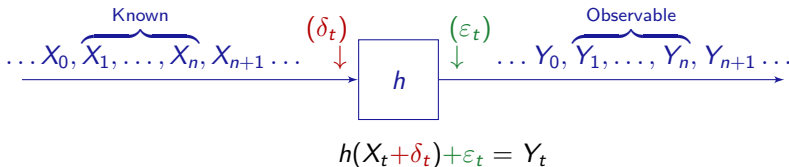
i.e., when risk-free outputs are in reasonable order

**Definition.** The outputs  $(Y_t)$  are **out of reasonable order** if

$$B_n := \frac{1}{n^{1/2}} \sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}| \rightarrow_{\mathbb{P}} \infty$$

...and then necessarily  $I_n \rightarrow_{\mathbb{P}} 0.5$

When do  $Y_t$ 's become out of reasonable order?



Roughly speaking, this happens when  $X_t$ 's are stationary and at least one of the risks does not vanish from  $I_n$ , that is, **when  $\delta_t$ 's and  $\varepsilon_t$ 's are**

- **noticeable** (e.g., if  $h(x) = c$ , then  $\delta_t$ 's vanish from  $I_n$ )
- **non-degenerate** (e.g., if  $\varepsilon_t = c$ , then they vanish from  $I_n$ )
- **not disguised as  $X_t$ 's** (e.g., if  $\delta_t = X_t$ ,  $\varepsilon_t = X_t$ , and  $h(x) = x$ , then  $Y_t = 3X_t$  and the risk-identifying “3” vanishes from  $I_n$ )

In the risk-free scenario, when do we have

$$I_n \rightarrow_{\mathbb{P}} \text{ somewhere } \neq 0.5 ?$$

- Let the stationary inputs  $(X_t)$  satisfy the **Glivenko-Cantelli** property and be **temperately dependent** (next two slides)
- Let  $h$  be almost everywhere differentiable with vanishing derivative outside an interval  $[a, b]$  (recall the AVR function)

Then in the risk-free scenario we have

$$I_n \rightarrow_{\mathbb{P}} I_{\infty} := \frac{\int_a^b (h'(x))_+ dx}{\int_a^b |h'(x)| dx} = \frac{1}{2} \left( 1 + \frac{\int_a^b h'(x) dx}{\int_a^b |h'(x)| dx} \right) \\ \neq 0.5 \quad \text{unless } h(b) = h(a)$$

**Definition.** Inputs  $(X_t)$  with the same marginal cdf's  $F$  satisfy the **Glivenko-Cantelli property** if  $X_1, \dots, X_n$  asymptotically identify  $F$ , that is,

$$\sup_{x \in \mathbb{R}} |F_n(x) - F(x)| \xrightarrow{\mathbb{P}} 0 \quad \text{when } n \rightarrow \infty$$

where  $F_n$  is the empirical cdf based on  $X_1, \dots, X_n$

- If the inputs  $(X_t)$  follow the **stationary ARMA( $p, q$ )** model driven by **iid innovations  $(\eta_n)$  with densities**, then they satisfy the **Glivenko-Cantelli property**

**Definition.** Inputs  $(X_t)$  with the same marginal cdf's  $F$  are **temperately dependent** if

$$\mathbb{P}(X_{1:n} \geq x) \rightarrow 0 \quad \text{and} \quad \mathbb{P}(X_{n:n} \leq x) \rightarrow 0$$

for all  $x$  such that  $F(x) \in (0, 1)$

## Examples

- If  $(X_t)$  are **iid**, then they are temperately dependent
- If  $X_t = X$  for all  $t \in \mathbb{Z}$ , then they are **not** temperately dependent (they are super-strongly dependent)
- If  $(X_t)$  are **strictly stationary and  $\alpha$ -mixing**, then they are temperately dependent

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## The system-monitoring index (statistic)

$$I_n = \frac{\sum_{i=2}^n (Y_{i,n} - Y_{i-1,n})_+}{\sum_{i=2}^n |Y_{i,n} - Y_{i-1,n}|} \rightarrow_{\mathbb{P}} I_{\infty} \begin{cases} = 0.5 & \text{risk affected} \\ \neq 0.5 & \text{risk free} \end{cases}$$

- is **simple** to implement
- **works** as intended in practically plausible situations
- jointly with another index, **helps to determine when** the system gets affected by risks: at the input, output, or both stages

A “stopping” question

When to sound the alarm?

$n=50?$   $100?$  ...

