

# The Inter-Arrival Time approach to Clustering

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## Agenda

The inter-arrival time approach to clustering

Clustering in empirical data: Atlantic Hurricanes

A model for clustering

Implications and further research

Appendix

# The inter-arrival time approach to clustering

- Intuitively, we characterise clustering by the following property: "An event makes the next one more probable."
- In general, (re)insurance contracts are sensitive to clustering of loss generating events.
- The degree of clustering affects risk assessment as well as risk capital allocation in a 1-year view.

# Risk estimation based on the distribution of annual losses

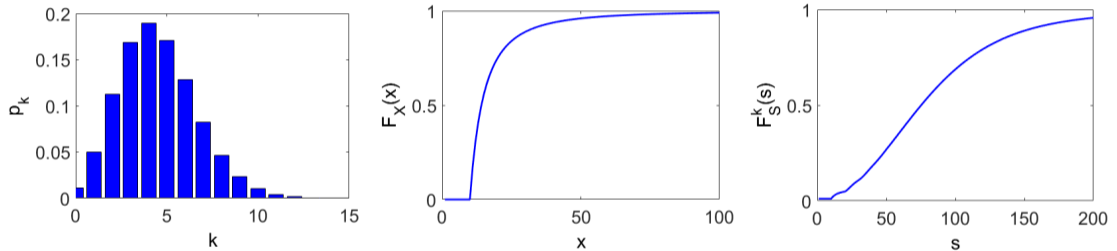


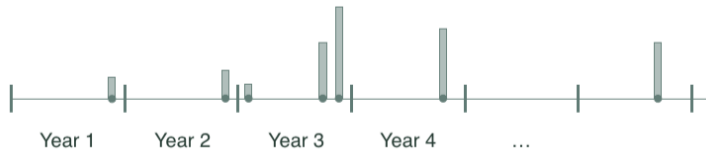
Figure: frequency distribution & severity distribution = annual loss distribution

# Explicit time makes the difference

Generate timestamps for events on the real time axis



Allocate a specific event loss to each timestamp



## How to construct a sequence of event times from empirical data?

- The IAT approach is based on inter-arrival times (IAT)  $\tau$ .
- The time series of events (arrivals) is generated from the empirical distribution of inter-arrival times  $\varphi_\tau$ .
- The process admits clustering if the distribution of IAT's is heavy-tailed.
- Let  $\tau$  be the time till the next event and  $t$  the time since the last. Then

" If  $\varphi_\tau$  is heavy – tailed, then  $E[\tau|t]$  increases in  $t$ ."  
(Sornette & Knopoff, 1997)

# Atlantic Hurricanes

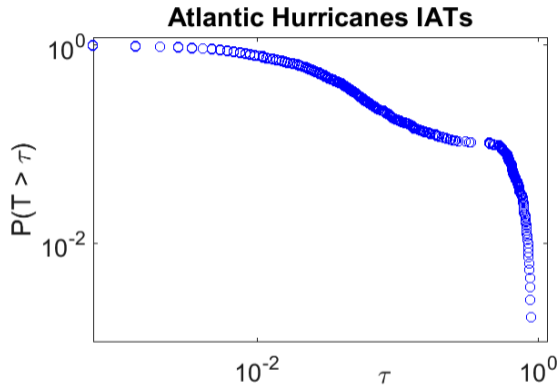
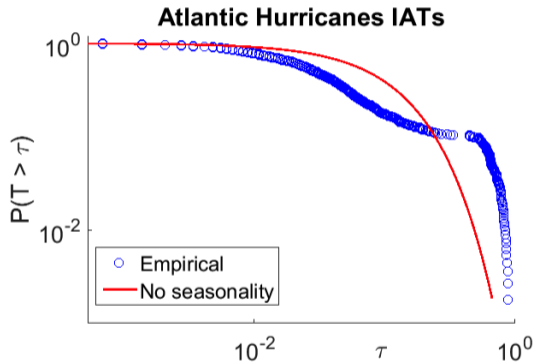
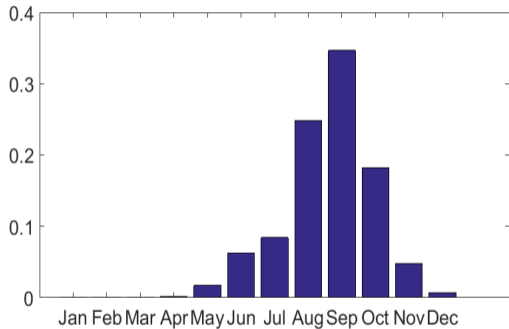
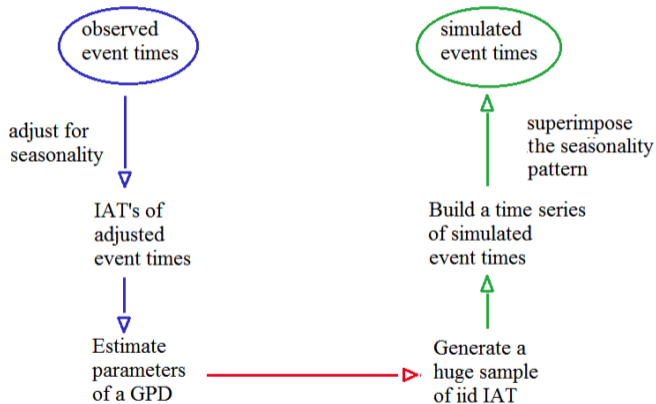


Figure: [HURDAT, 1891 - 2008]

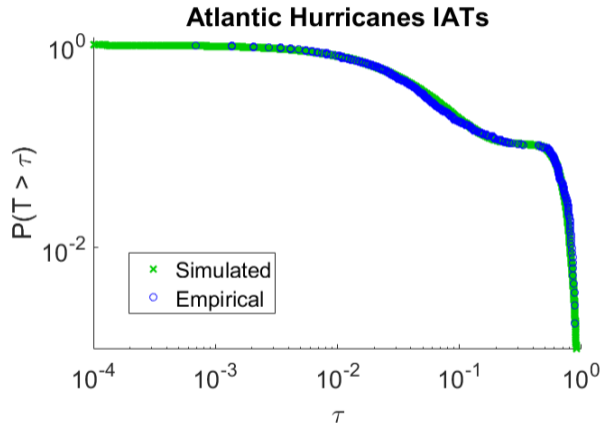
# Adjusting for seasonality



# From empirical to modelled IATs



# Modelled data matches empirical data



## The general Pareto distribution is a generic model

- Empirical data provides statistical evidence that the *generalised Pareto Distribution* (gPD) is a superior model.

$$\text{if } \tau \sim \text{Exp}(\lambda) \mid \lambda \sim \Gamma(\alpha, \beta) \text{ then } \tau \sim \text{gPD}(\xi, \sigma, 0).$$

- shape parameter  $\xi = 1/\alpha$ ,  $0 \leq \xi < 1$ ;
- scale parameter  $0 < \sigma = \beta/\alpha$ .

## Implied properties

- *"The longer it has been since the last event 't' the longer is the expected time  $\tau$  till the next."*
- The gPD admits clustering

$$E[\tau|t] = \frac{\sigma}{1-\xi} + t \frac{\xi}{1-\xi}$$

- The gPD admits over-dispersion due to

$$\delta_{\xi} := \frac{\text{VAR}(\tau)}{\text{MEAN}(\tau)^2} = \frac{1}{1-2\xi} \geq 1, \quad 0 \leq \xi < \frac{1}{2}$$

## Possible model extensions

- Introduce truncation of IAT distribution
  - Lower bound due to temporal resolution in empirical data
  - Upper bound, e.g. due to periodicity
- Include peril specific external drivers
  - Atlantic Multidecadal Oscillation, El-Niño Southern Oscillation
  - Business activity cycle
  - Trends

# The negBin as a frequency model

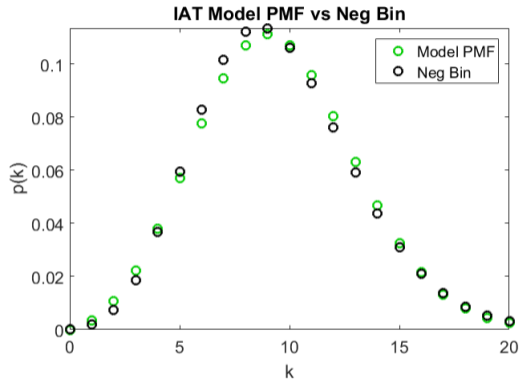


Figure: The PMF of the 1-year truncated gPD model approximates a negBin.

## Memory: Some implications

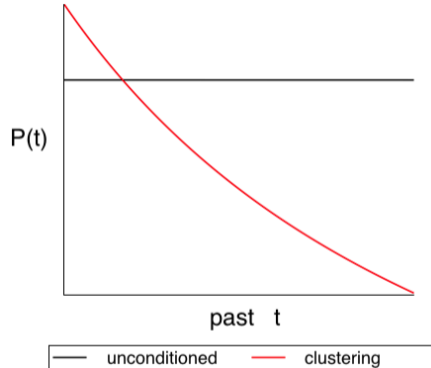
- Risk assessment based on samples drawn from 1 year disregards inter-year correlation.
- The economic 1 year view imposes an external time scale.



- Should the 1-year view on risk be conditioned on the past?

$$P(t) = E[\tau \leq 1 \mid t]$$

# Conditioning on the past alters expectations about future arrivals



## Further research

- Relation between the IAT approach and cluster analysis
- Scale dependence: In contrast to Poisson processes, processes with memory are susceptible to externally imposed time scales. Thus, results in general depend on the observation period.
- Inter-arrival times might depend on event severities: "Do strong storm cluster more than weak ones?"
- Multiperils: How to describe inter-dependencies between different perils in the IAT frame work?

**Thank you very much for your attention!**

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## References



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