



# Stochastic Modeling of Daily Rainfall Process in a Non-Stationary Context

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*Project 1-4*

NSERC Canadian FloodNet 3<sup>rd</sup> Annual General Meeting

June 27- 28, 2017

# Outline

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- **Objective:** Stochastic Modeling of Extreme Rainfall Events

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- **Methodology:** Stochastic Approach: MCME model
    - **Results**

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- **Future work:**
    - **Linking with Climate Variables**

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- **Conclusion**

# Objective

## Stochastic Modeling of Daily Rainfall Processes in a Non-Stationary Context

- **Stochastic Approach**
  - Another modelling technique besides the statistical approach
- Will take full daily rainfall series as input
  - Will generate accurate synthetic series



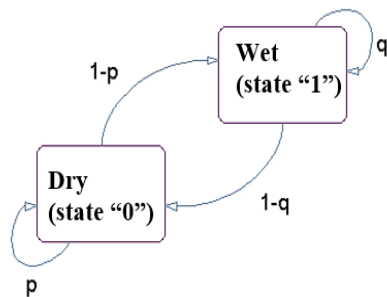
# METHODOLOGY

Stochastic Approach: MCME model

# MCME model

- MCME = Markov Chain Mixed Exponential
- 2 components:
  - 1) **Rainfall Occurrence:** Markov chain
  - 2) **Rainfall Amount:** Mixed Exponential Distribution

1<sup>st</sup> order 2-state Markov Chain



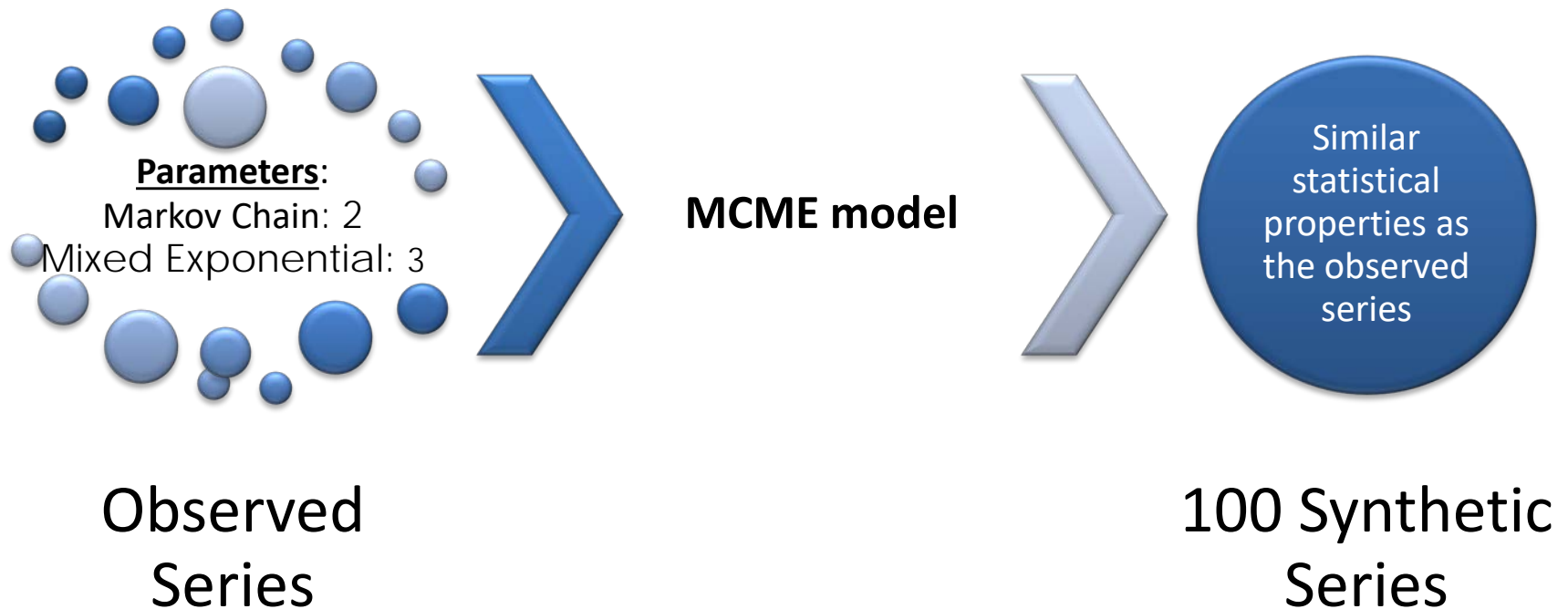
## *Mixed Exponential Distribution*

$$f(x) = \left(\frac{p}{\mu_1}\right) e^{-x/\mu_1} + \left(\frac{1-p}{\mu_2}\right) e^{-x/\mu_2}$$

$$\begin{cases} x > 0 \\ p: \text{mixing probability}; 0 < p < 1 \\ \mu_1: \text{scale parameter}; 0 < \mu_1 < 30 \\ \mu_2: \text{scale parameter}; 0 < \mu_1 < \mu_2 \end{cases}$$

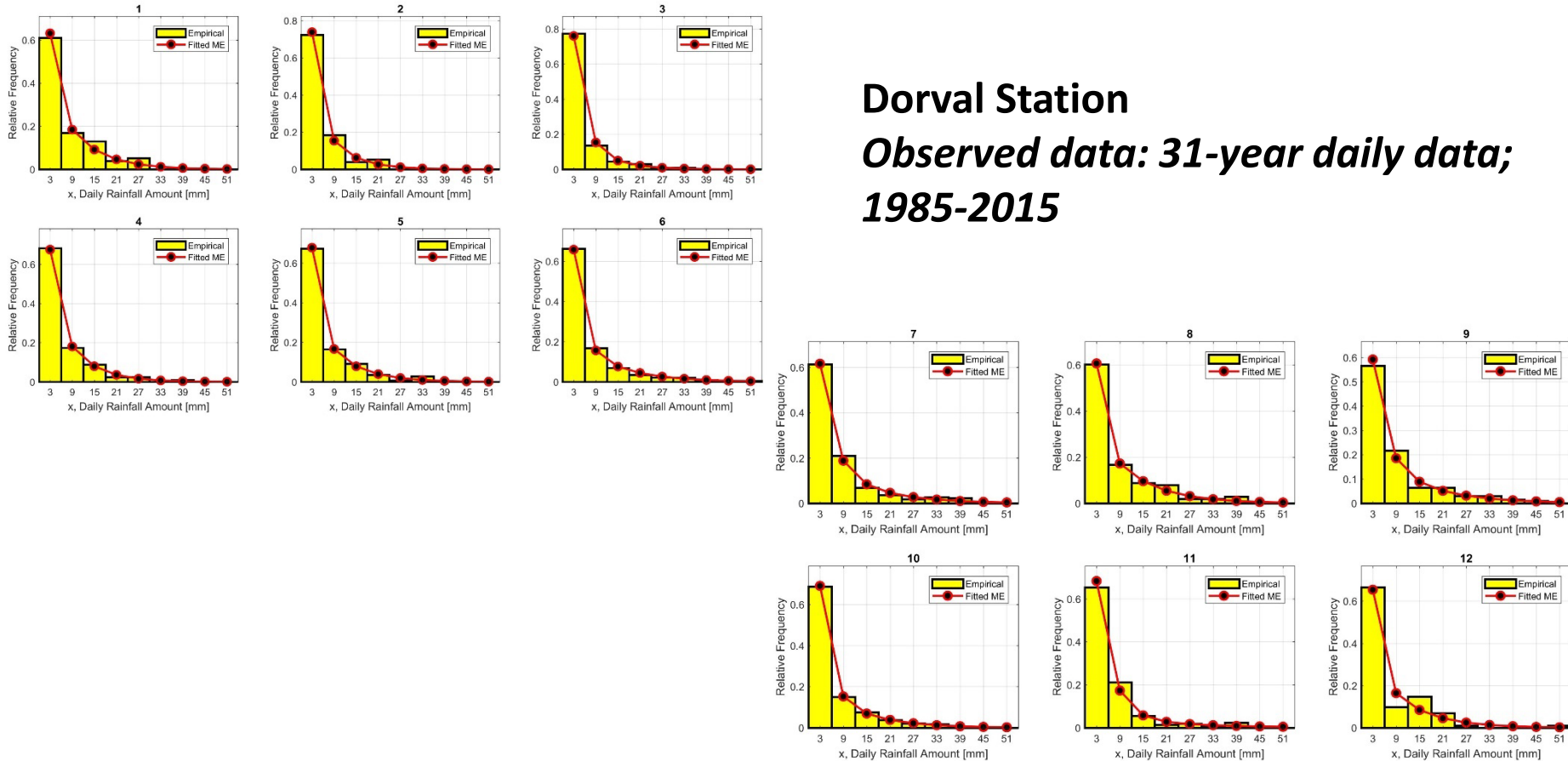
- Seasonality variation accounted by **Fourier Series**
- The **Shuffled Complex Evolution (SCE)** technique was used for global optimization of the maximum likelihood

# MCME model



# Results

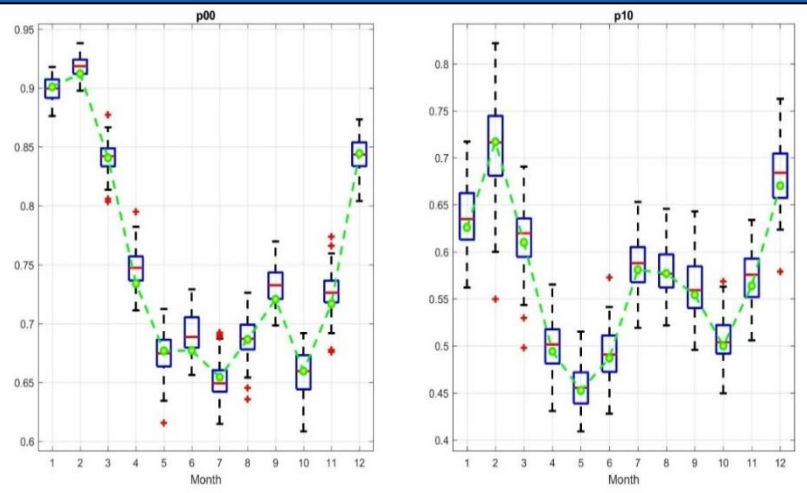
## Monthly histograms of the observed series with the fitted theoretical mixed exponential distribution



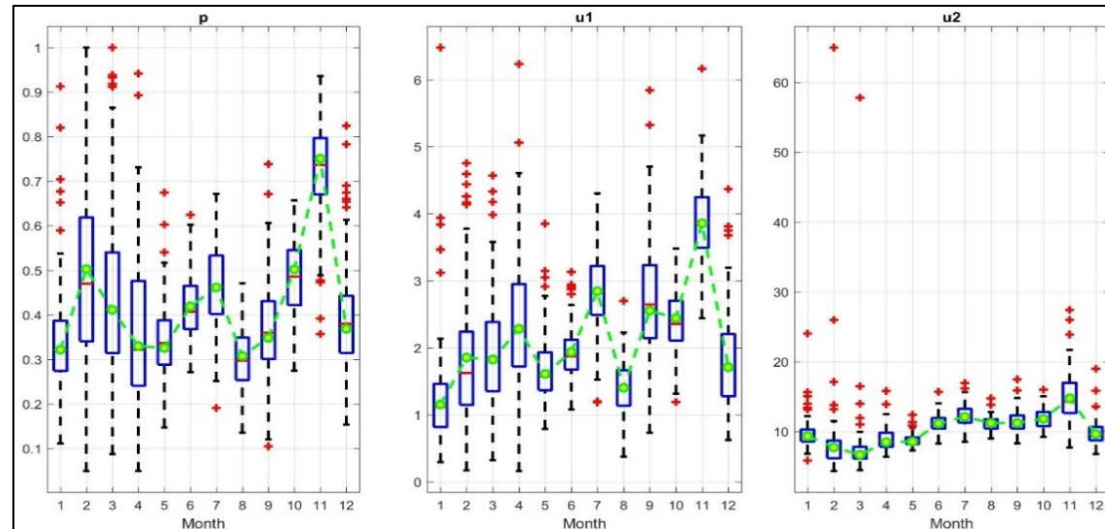
# Results

## Monthly boxplots of parameters variation

*Markov chain parameters*



*Mixed Exponential parameters*





# MCME model

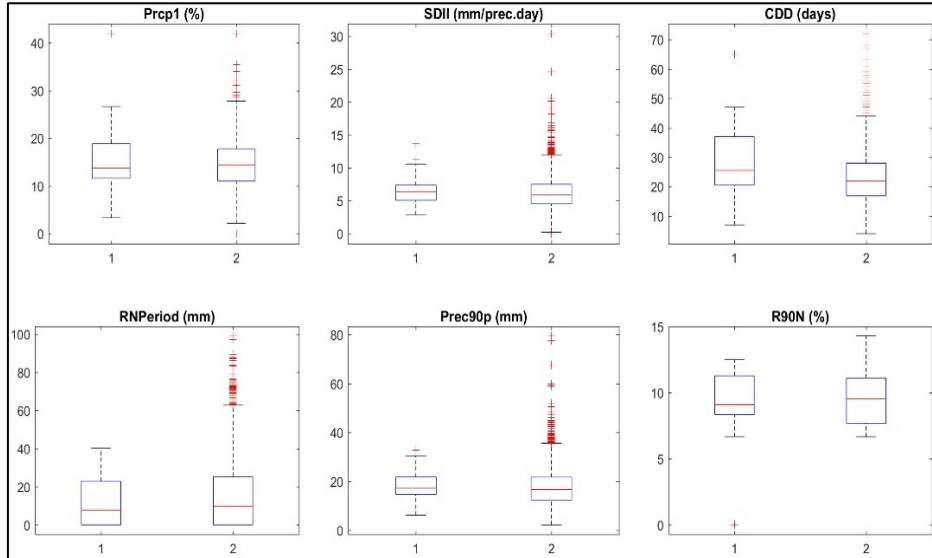
## Model Evaluation

| Rainfall Property | Index   | Definition  | Unit |
|-------------------|---------|---|------|
| Frequency         | Prcp1   | Percentage of wet days<br>(Threshold $\geq 1$ mm)             | %    |
| Intensity         | SDII    | Mean precipitation<br>amount at wet days                      | mm   |
| Extremes          | CDD     | Maximum number of<br>consecutive dry days                     | days |
|                   | R3Days  | Maximum 3-days<br>precipitation total                         | mm   |
|                   | Prec90p | 90 <sup>th</sup> percentile of rain<br>day amount             | mm   |
|                   | R90N    | % days with<br>precipitation > 90 <sup>th</sup><br>percentile | %    |

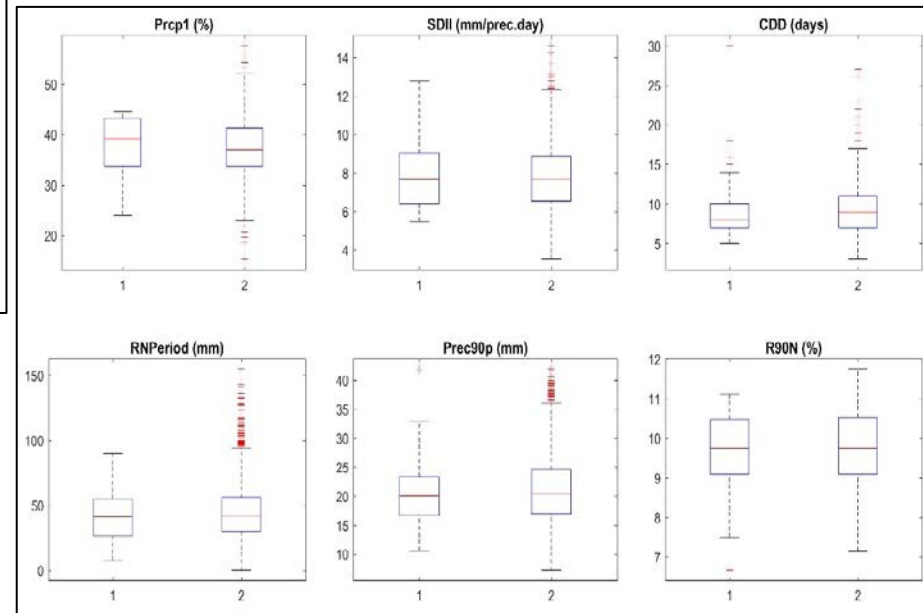
(Gachon et al., 2005)

# Results

## Model Evaluation



*Winter indices- 1: Observed – 2: Simulated*



*Summer indices- 1: Observed – 2: Simulated*



# FUTURE WORK

Linking with Climate Variables

# Future Work

- One statistical downscaling technique:  
**Stochastic Weather Generators**
  - Example: **LARS-WG**
    - Generates synthetic daily time series of climate variables, statistically identical to the observed series, taking into account  $\Delta$ change between the baseline period and the selected future period
- **Future Challenges:**
  - Linking MCME with Global Climate Models
  - Comparing with LARS-WG
  - But, how to account for **non-stationarity**??

# Conclusion

- Main Objective:
  - New Methodologies for Updating IDF curves
- The **stochastic approach** is adopted
- The MCME is developed to generate synthetic rainfall series statistically identical to the observed ones
- Including the **climate variables** in the MCME model will help taking into consideration **the climate change**



# THANK YOU!

Questions?