

Extreme Rainfall in a Multi-scale Modeling Framework

Morgan Phillips, Scott Denning, Mazdak Arabi, Mark
Branson

CMMAP Team Meeting

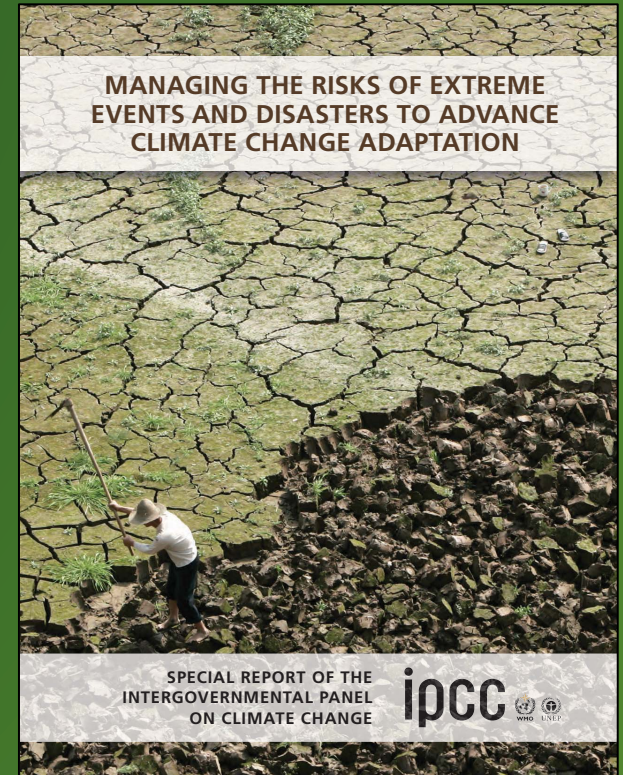
August 4, 2015



Extreme Precipitation: Why do we care?

- Extreme rainfall: Occurrence of rainfall near the upper ends of observed values
 - Damage to assets and infrastructure, loss of life
 - Physical interaction of rainfall with land surface

How might extremes in precipitation behave under changing climate?



Extreme Precipitation: Scale Dependence



10 mm/hr

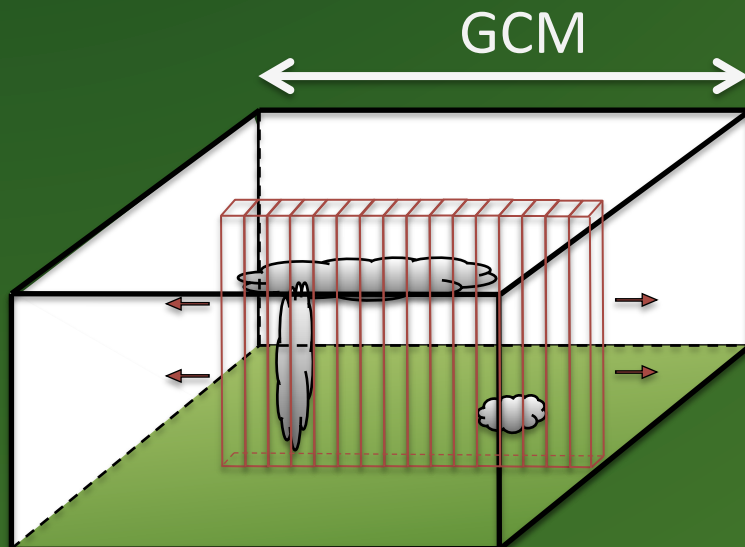
2 mm/hr

Precipitation intensity is function of spatial resolution

Introduction: Downscaling

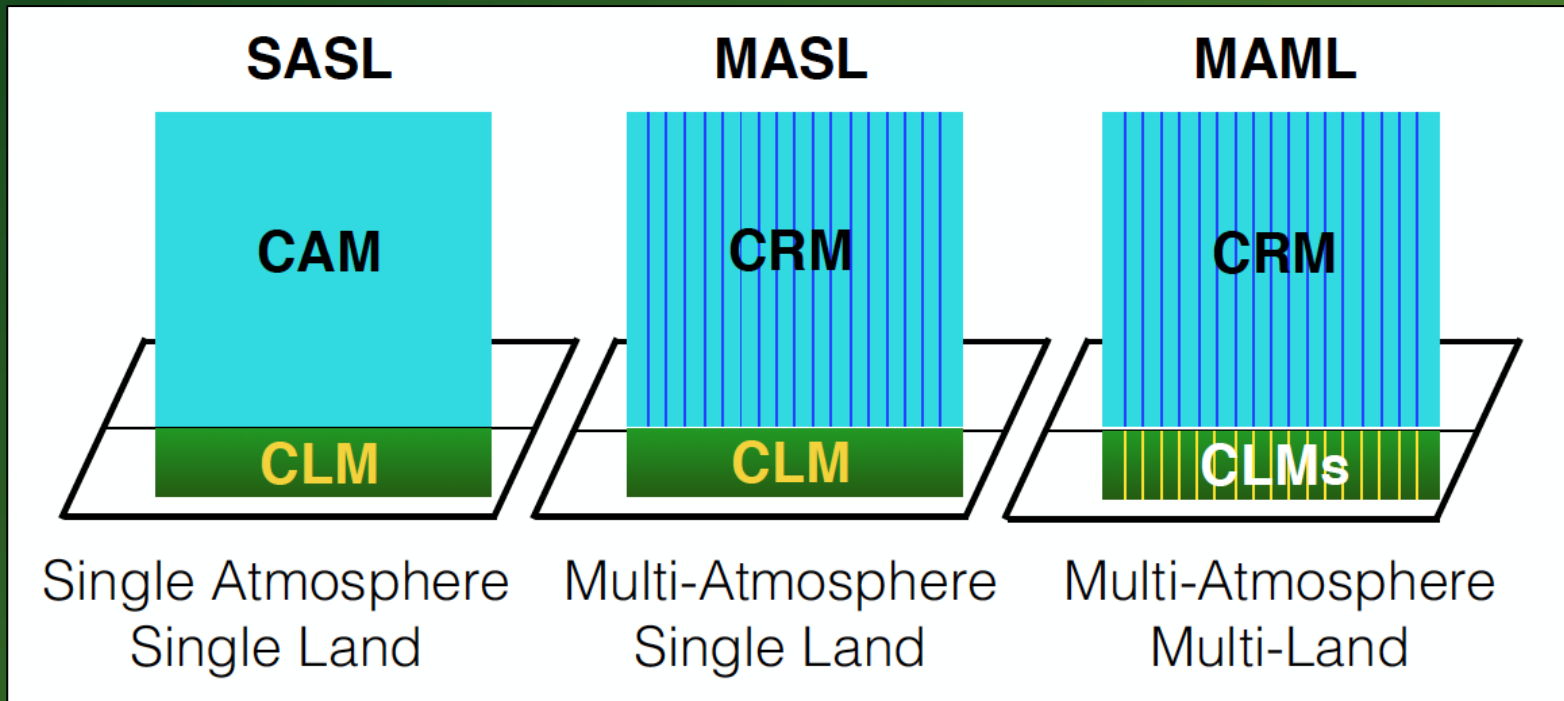
Downscaling techniques

- Statistical
 - Develop relationship between observed quantity and Global Climate Model (GCM) quantity
- Dynamical
 - GCM quantities used to force regional mesoscale model



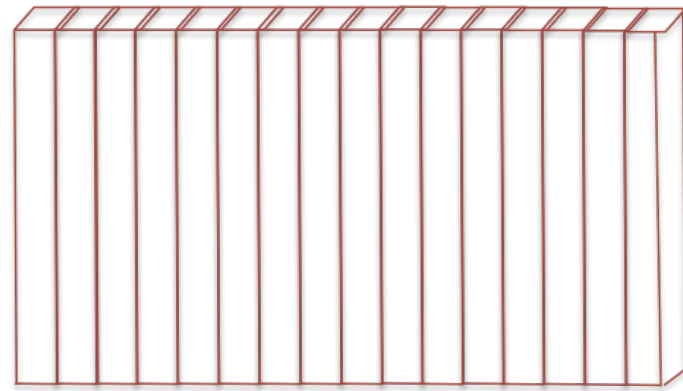
- Multi-scale Model Framework (MMF)
 - Cloud Resolving Model (CRM): Information about atmospheric structure on O[1 km]
 - MMF correctly simulates intense rainfall events compared to traditional GCM [DeMott et al., 2007]

Introduction: Land Atmosphere Coupling



Introduction: Flux at Land-Atmosphere Interface

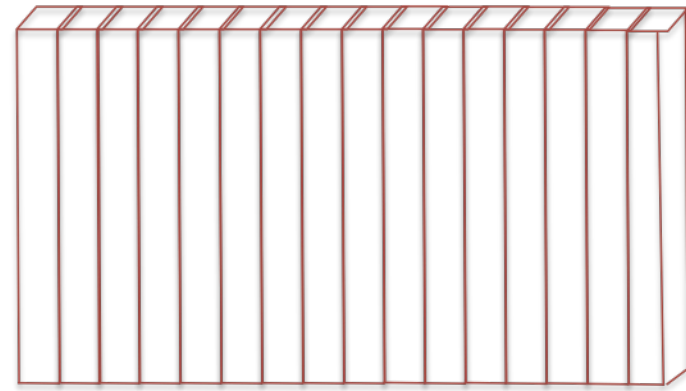
MASL



$\overline{F_{Atmos}}$



MAML



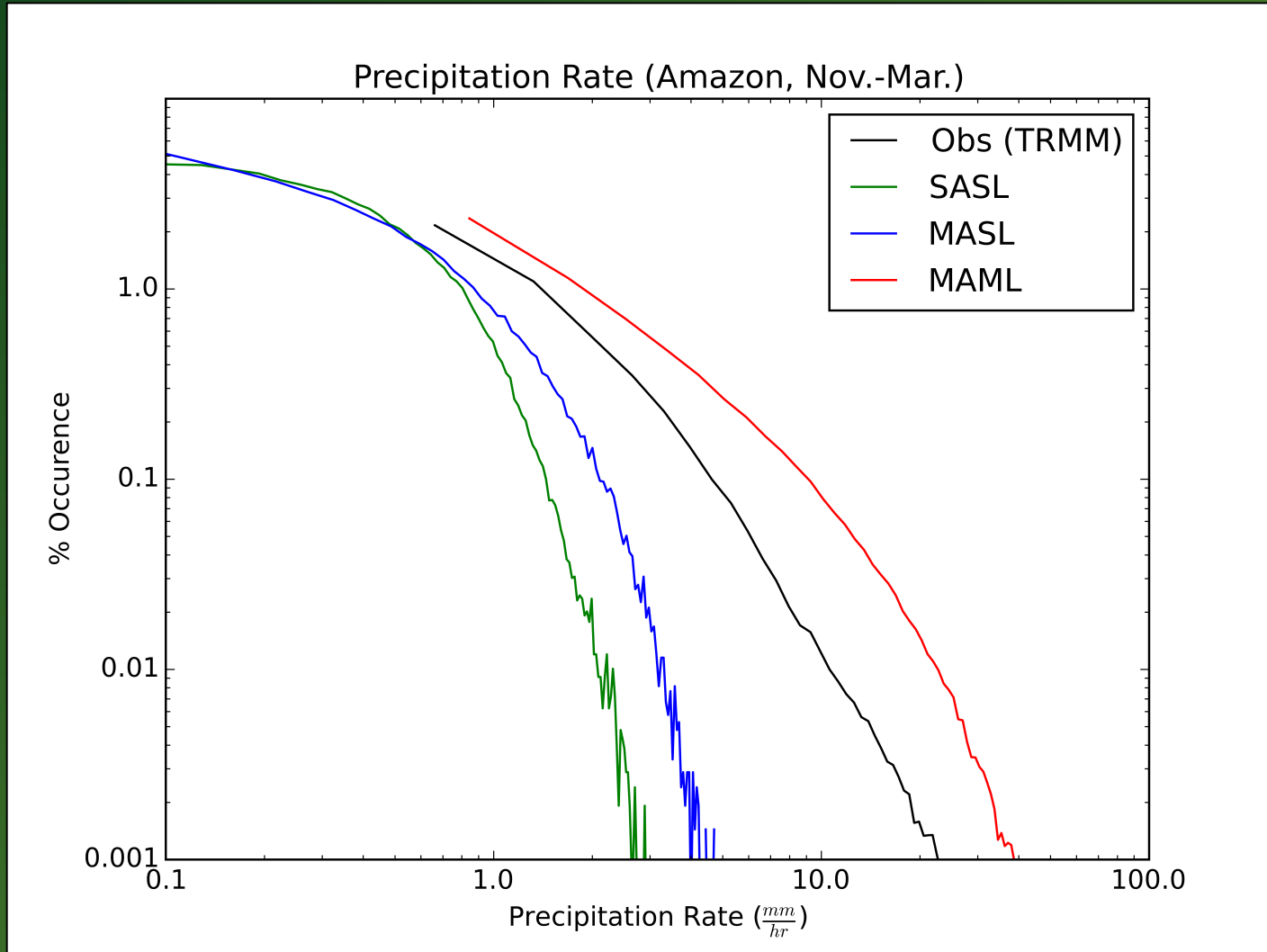
F_{Atmos_i} $F_{Atmos_{i+1}}$...



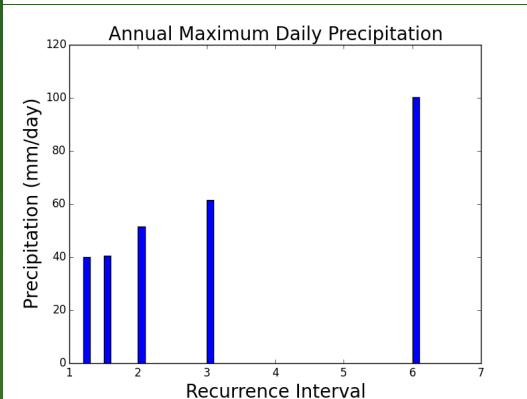
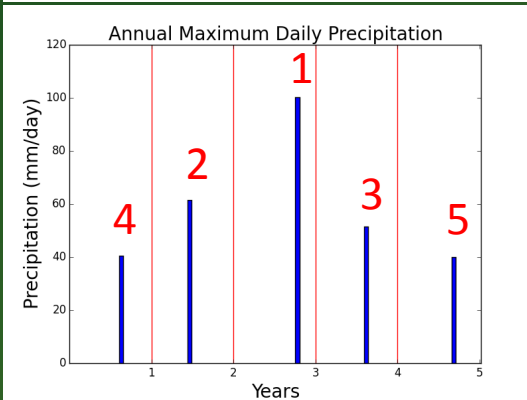
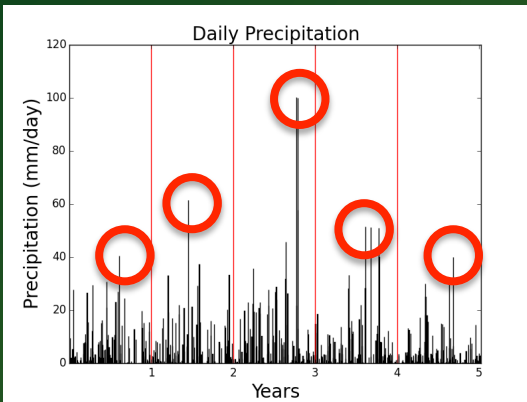
$F_{Atmos_{i+n}}$



Introduction: Precipitation Intensity at Land-Atmosphere Interface



Methods: Annual Maximum Series



Select largest annual precipitation from time-series

Rank by magnitude

Express in terms of Recurrence Interval (T) (Gumbel, 1945)

$$T = \frac{1+n}{m}$$

$$n = \# \text{ years}$$
$$m = \text{rank}$$

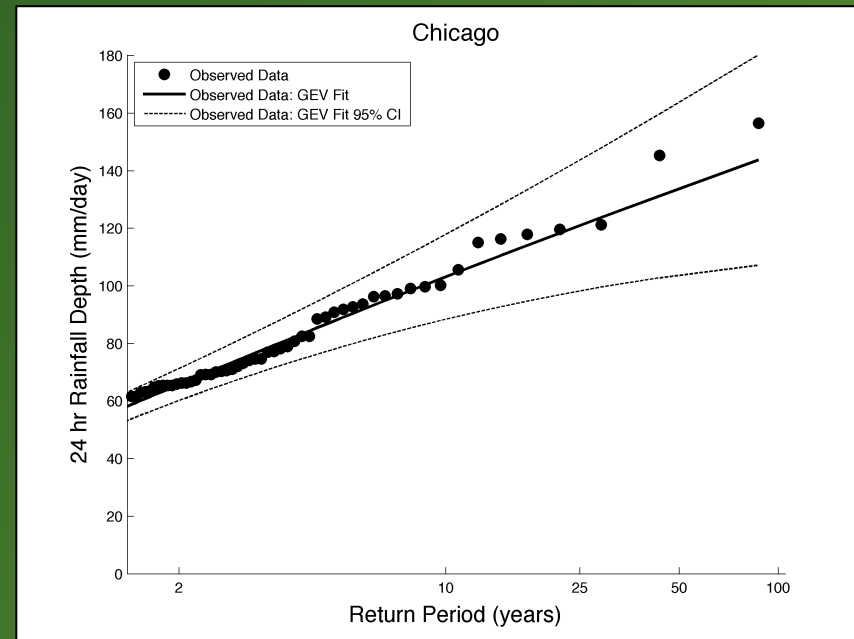
$$P = \frac{1}{T}$$

Methods: Annual Maximum Series

- Distribution follows extreme value type [Gumbel, 1941 and Chow, 1951]
- General Extreme Value distribution best suited for meteorological variables [Jenkinson, 1955]

General Extreme Value Distribution

- Estimate 3 parameters: Shape, Scale and Location [Jenkinson, 1969]
- Evaluate uncertainty in parameter estimates using confidence intervals

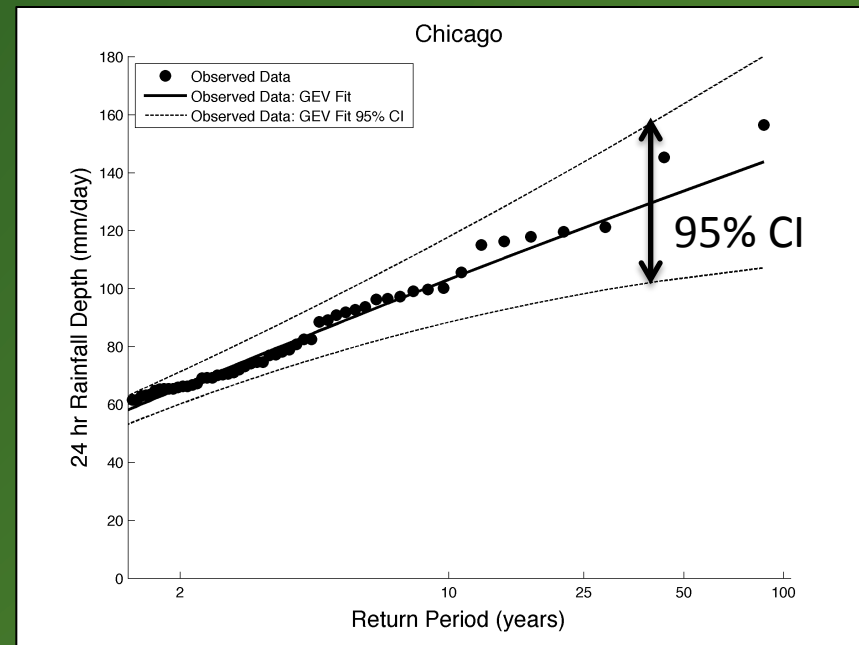


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General Extreme Value Distribution

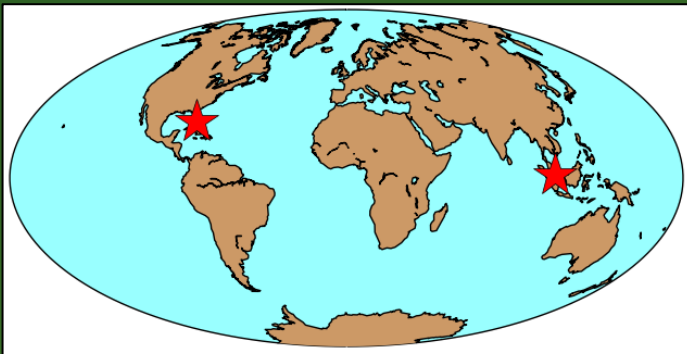
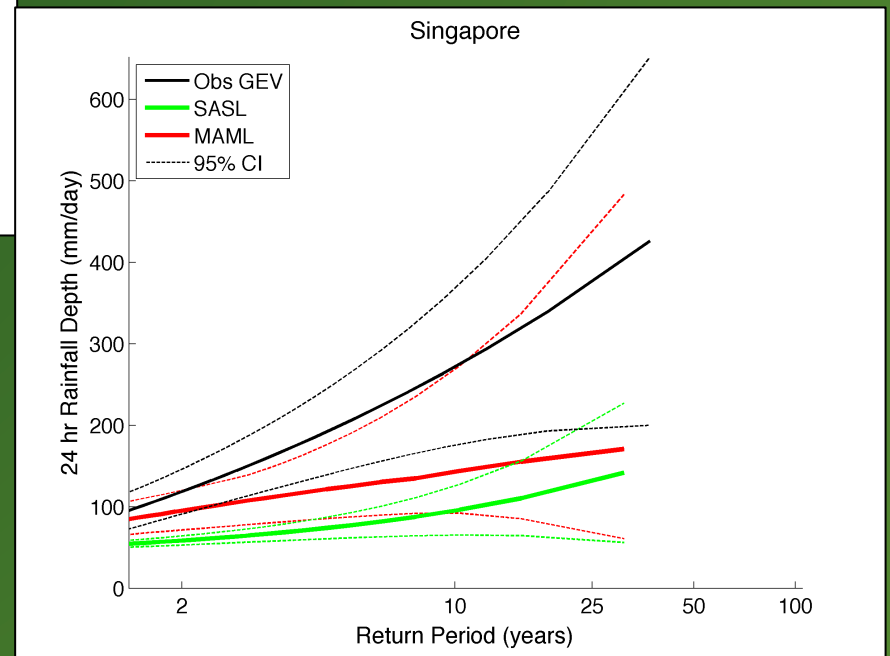
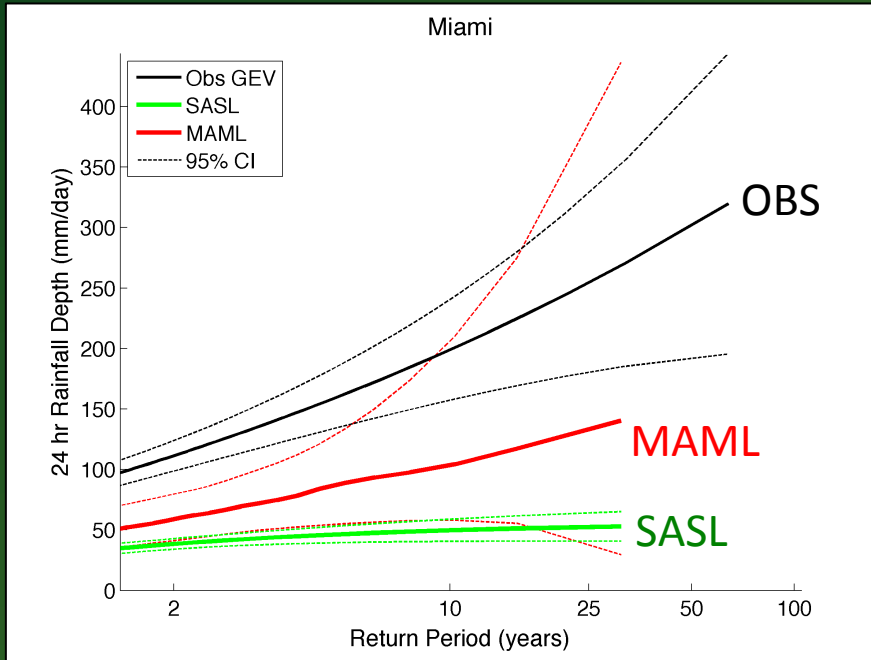
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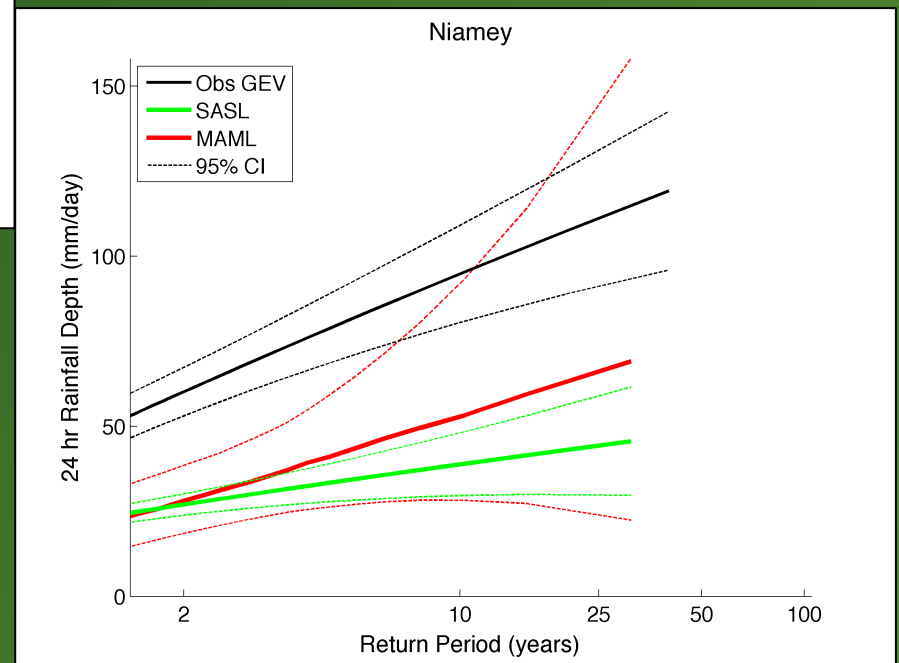
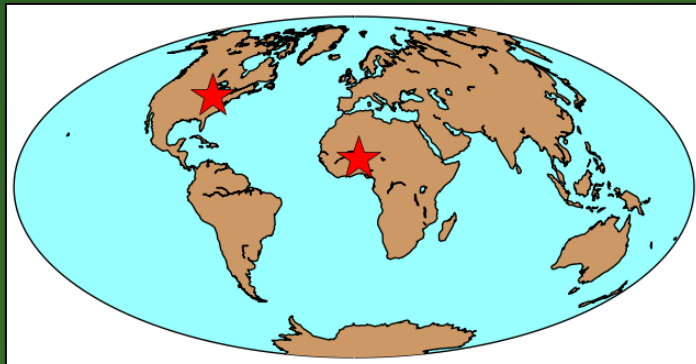
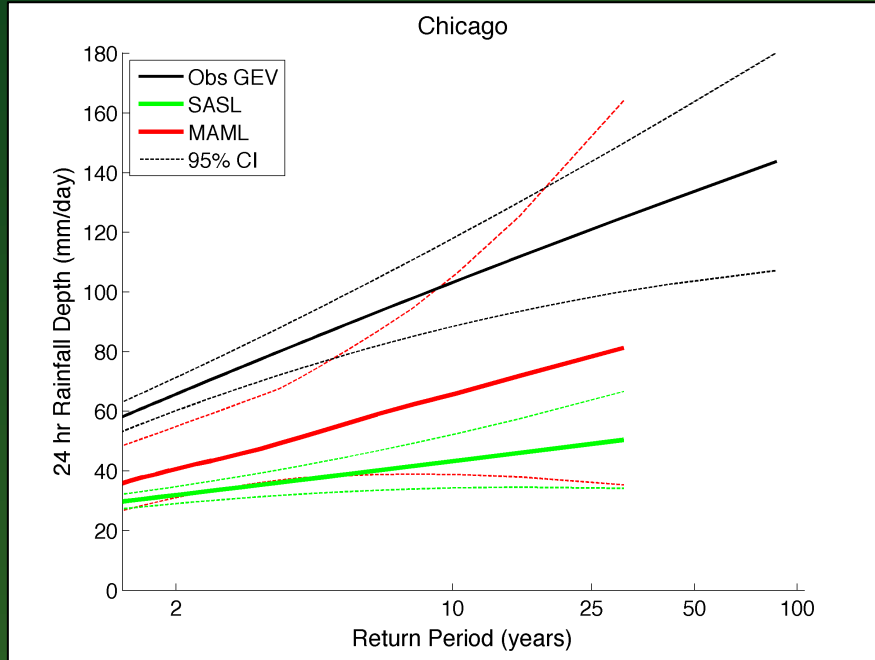
Methods: Model Experiment

- Community Climate System Model
 - Prescribed ocean/ice, interactive land
- SASL, MASL and MAML simulations
 - 30 years integration with precipitation history output @ 24 hours
 - Climatological SST (fixed monthly average)
 - GCM resolution of 1.9 x 2.5 degree
- MASL and MAML
 - CRM resolution of 32 x 4 km

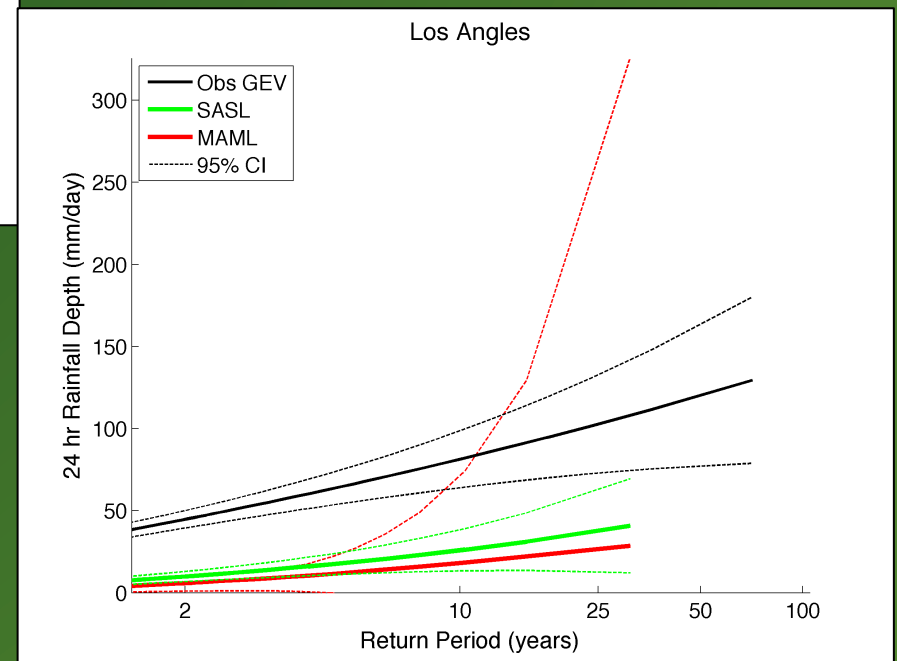
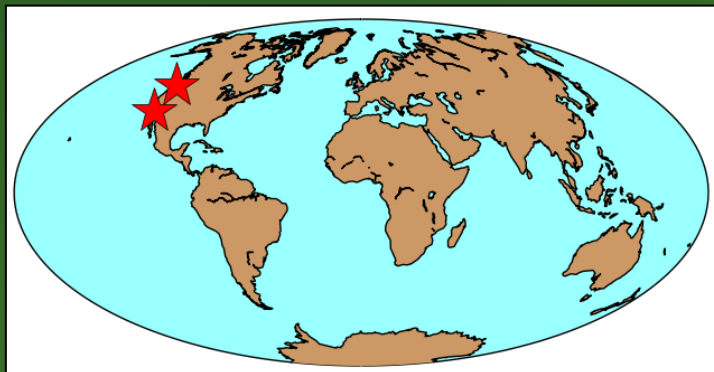
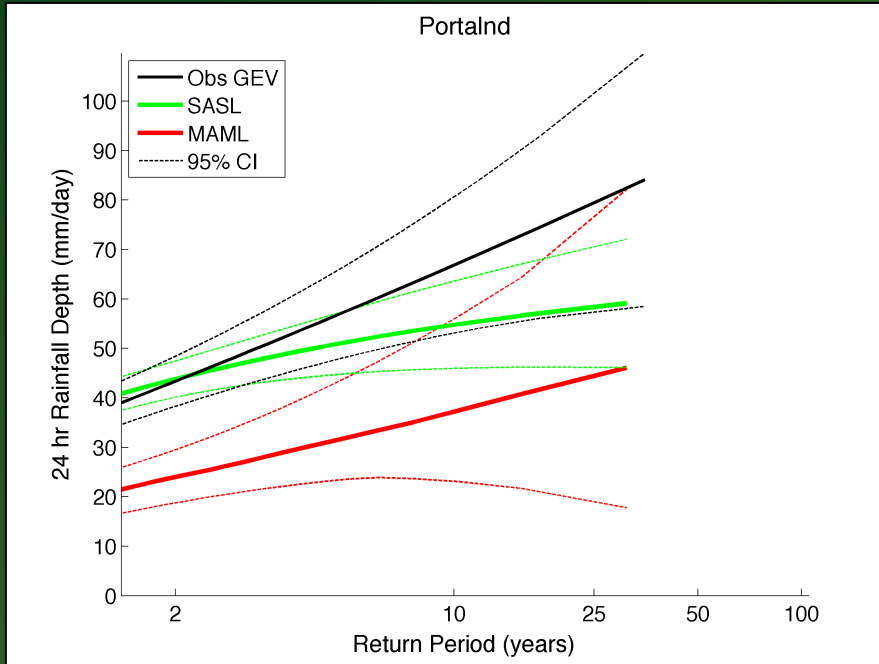
Results: Coastal Convective



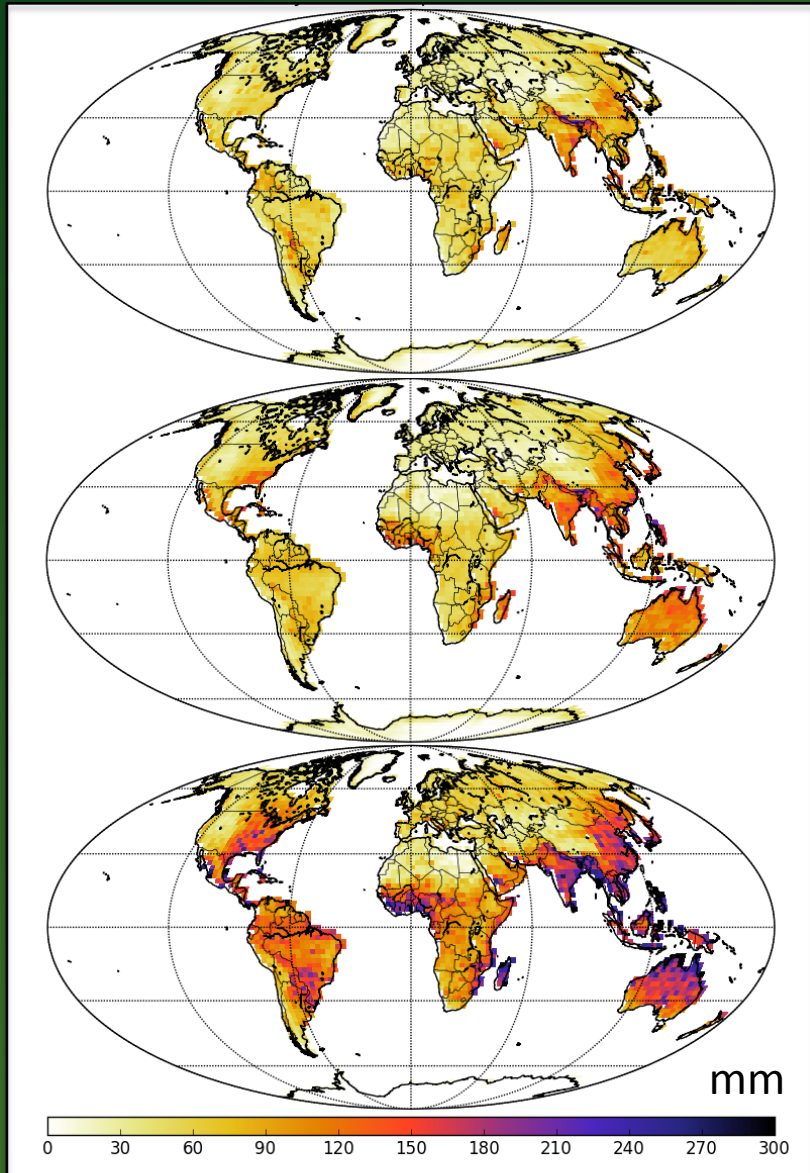
Results: Continental Convective



Results: Coastal Stratiform



Results: 30 Year, 24-Hour Storm Depth



SASL

MASL

MAML

Future Work

- Quantitative measure to evaluate simulated distributions against observations
- Global evaluation of model statistics against observations
- Investigate regions where MMF extremes are more like observations
- Sub-daily statistics and climate variability

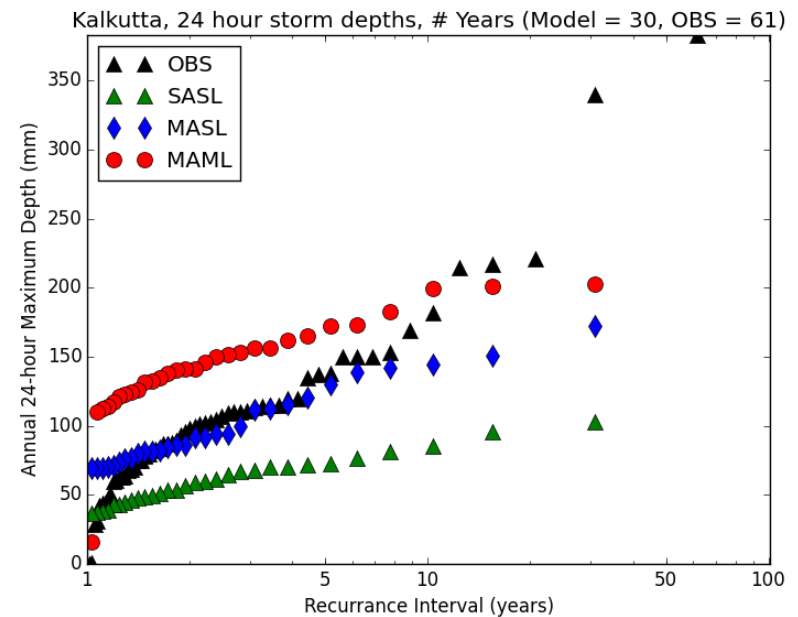
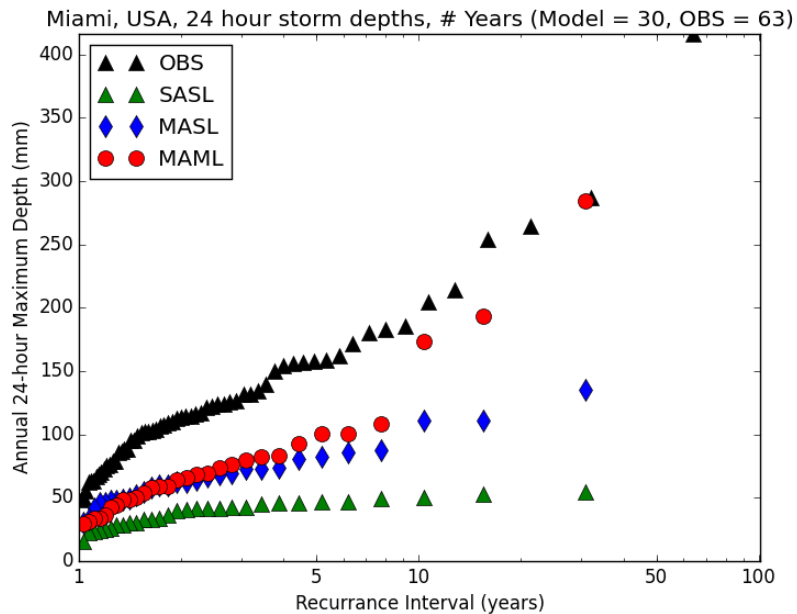
References

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Acknowledgements

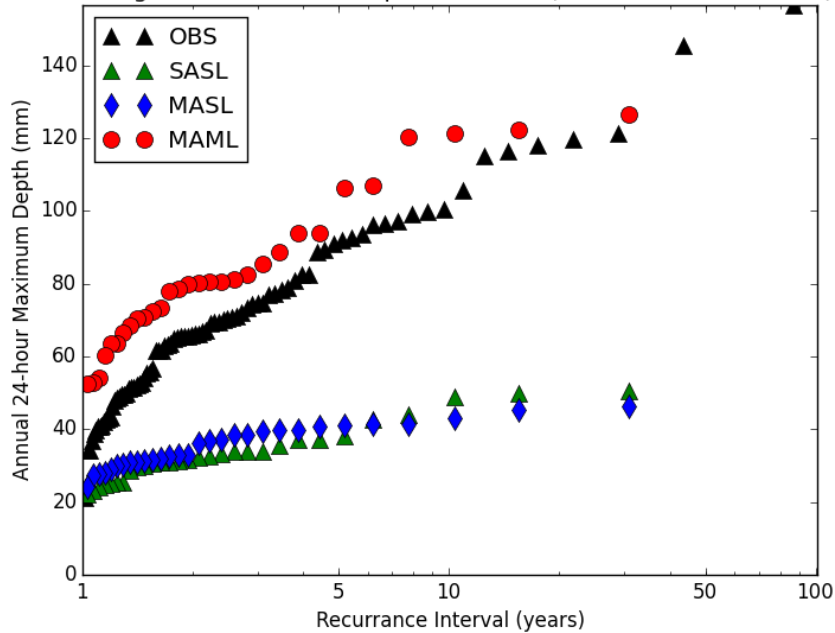
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Appendix: SASL-MASL-MAML

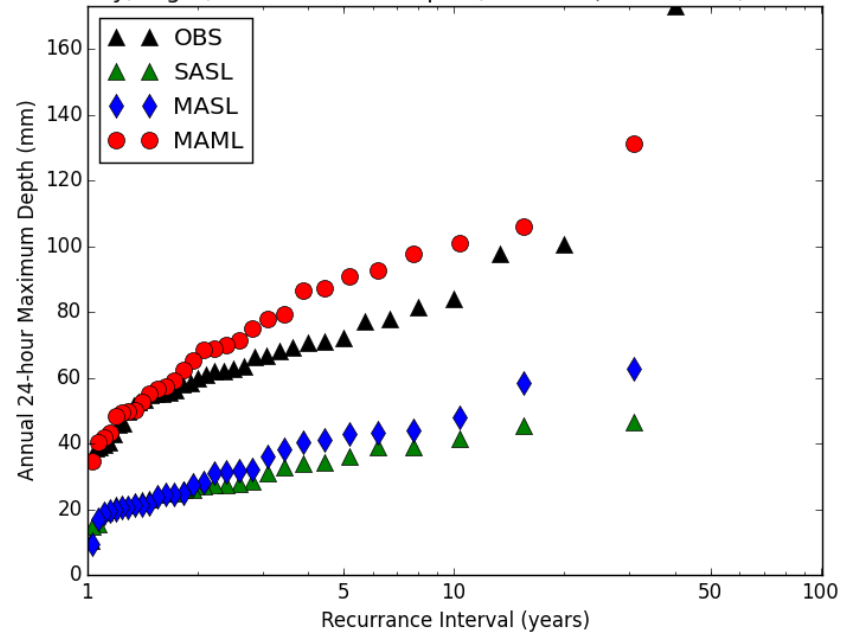


Appendix: SASL-MASL-MAML

Chicago, 24 hour storm depths, # Years (Model = 30, OBS = 86)

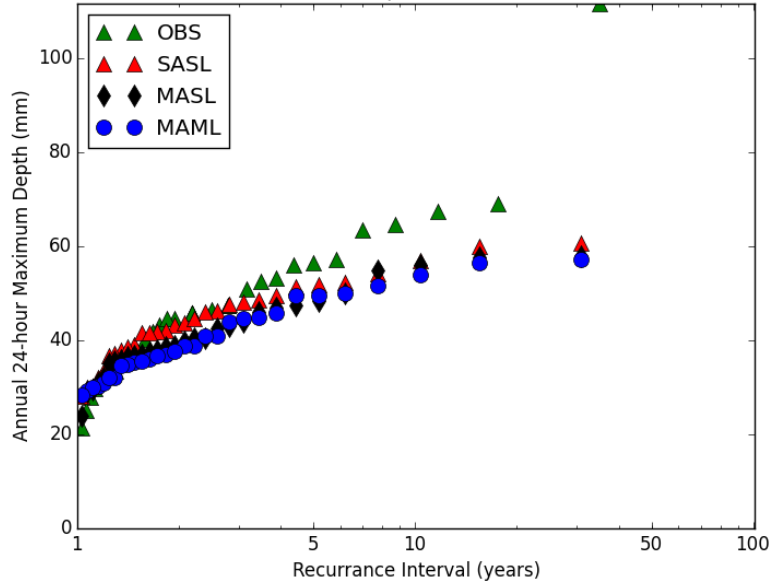


Niamey, Niger, 24 hour storm depths, # Years (Model = 30, OBS = 39)

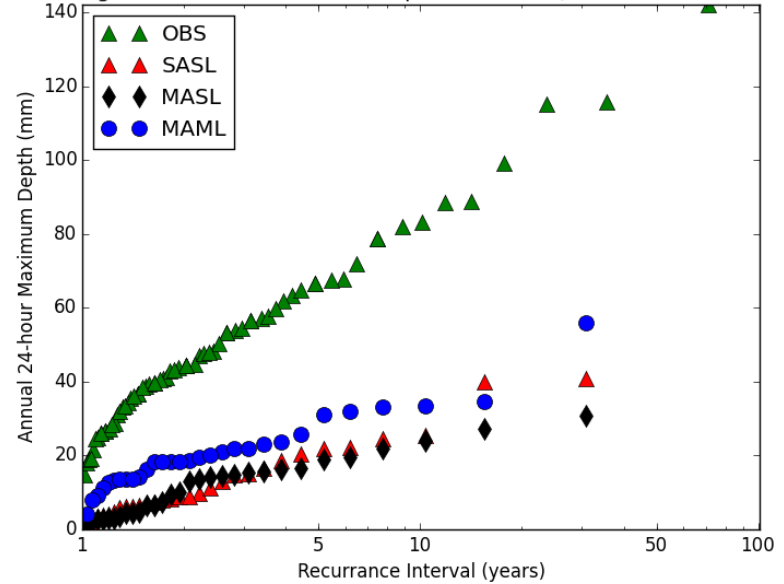


Appendix: SASL-MASL-MAML

Portland, USA, 24 hour storm depths, # Years (Model = 30, OBS = 34)



Los Angeles, USA, 24 hour storm depths, # Years (Model = 30, OBS = 70)



Outline

- Introduction
 - Extreme Rainfall in traditional Global Climate Models
 - Why do we care
 - Current methods to evaluate extreme rainfall from climate models (statistical vs dynamical)
 - MAML, MASL and SASL model architecture
- Methods/Results
 - Annual Maximum Series
 - Extreme Value Fitting
 - Spatial Trends
 - Future Work

Extreme Value Fitting: General Extreme Value Distribution

$$f(x) = \frac{1}{\sigma} \left[1 + \zeta \left(\frac{x - \mu}{\sigma} \right) \right]^{-\frac{1}{\zeta} - 1} \exp \left\{ - \left[1 + \zeta \left(\frac{x - \mu}{\sigma} \right) \right]^{-\frac{1}{\zeta}} \right\}$$

General Extreme Value Distribution (GEV)
Cumulative Density Function
 ζ : Shape σ : Scale μ : Location

$$F(x) = \exp - \left[1 + \zeta \left(1 + \left(\frac{x - \mu}{\sigma} \right) \right)^{-\frac{1}{\zeta}} \right]$$

GEV: Quantile Function

$$x = \mu + \frac{\sigma}{\mu} \left[-1 - \ln F(x) \right]^{-\zeta}$$

ζ, σ, μ estimated by maximizing the log-likelihood function of the GEV

$$L(\mu, \sigma) = - \sum_{i=1}^n \frac{x_i - \mu}{\sigma} - n \ln \sigma - \sum_{i=1}^n \exp \left[- \frac{x_i - \mu}{\sigma} \right]$$

$$\{ \mu, \sigma \in \mathfrak{R} \mid \frac{\partial \ln(\mu, \sigma)}{\partial \mu} = \frac{\partial \ln(\mu, \sigma)}{\partial \sigma} = 0 \}$$