# Monsoon Dynamics and Computational Modeling Techniques

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## Outline



### 1 Why Study Monsoons?

- What Defines a Monsoon? Monsoon Intraseasonal Oscillations
- 3 Numerical Weather Prediction Euler Equations and Fluid Dynamics Advanced Research WRF and HPC
- Novel Computational and Numerical Techniques Benefits of Parallelization Radial Basis Function-generated Finite Difference Methods

## 5 Summary



- Long history of inaccurate forecasts<sup>6</sup>.
  - Incorrect forecasts cause speculation.
- Monsoon rains intrinsically impact Indian, and global, economy<sup>3</sup>.
  - Normal monsoon 96-104% of 89cm 50-year average each season.
  - Below 90% = drought.
  - Vital for agriculture sector 15% of \$2 trillion econonmy.
  - Monsoon boosts market demand and can mitigate inflation.
- Climate Disasters
  - 2015 Chennai Floods
  - 2016 Drought and Mass Migration<sup>1</sup>
- Better Prediction = less speculation and more resiliency



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However, this description does not account for certain monsoon features, like delayed onset or the active-break cycle.





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The Climate of Tropical Regions. URL: http://thebritishgeographer.weebly.com/the-climate-oftropical-regions.html Historically, Halley (1686) attributed monsoons to heating contrast between land and ocean, i.e. an enormous sea-breeze[4].

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#### Low Frequency MISO (30-60 Day) - Northward propogating



#### High Frequency MISO (10-20 Day) - Westward propogating



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Performing NWP involves solving Navier-Stokes equations, simplified as the Euler Equations:

$$\begin{aligned} \partial_t U + (\nabla \cdot \mathbf{V}u) &- \partial_x (p\phi_\eta) + \partial_\eta (p\phi_x) = F_U \\ \partial_t V + (\nabla \cdot \mathbf{V}v) &- \partial_y (p\phi_\eta) + \partial_\eta (p\phi_y) = F_V \\ \partial_t W + (\nabla \cdot \mathbf{V}w) - g(\partial_\eta p - \mu) = F_W \\ \partial_t \Theta + (\nabla \cdot \mathbf{V}\theta) = F_\Theta \\ \partial_t \mu + (\nabla \cdot \mathbf{V}) = 0 \\ \partial_t \phi + \mu^{-1} [(\mathbf{V} \cdot \nabla \phi) - gW] = 0 \end{aligned}$$

This tracks momentum in 3D, potential temperature conversation, mass conservation, and geopotential respectively.

 $F_{\star}$  terms represent forcings from model physics, turbelent mixing, spherical projections, Coriolis forces, etc. In WRF, options select physics suites and orders of accuracy.



The Weather and Research Forecasting (WRF) model solves the Euler governing equations in separate horizontal and vertical domains via a Finite Volume solver using explicit Finite Difference approximations.





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Weighted Essentially Non-Oscillatorry (WENO) schemes comprise flux formulations

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## Air-Sea Coupling



WRF alone poorly models ISM. So we seek to couple to an ocean model. Sea Surface Temperatures and D20 isotherm depths are shown below.



# Air-Sea Coupling



Averaged rainfall Jun-Jul-Aug-Sep are shown where surface wind stress, net energy fluxes, and SST are exchanged in coupled model (Samala).



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## NWP and Communication Overhead



WRF simulations can take days to run depending on parameters, so speedup is crucial!

### MPI Only Excessive halo communication



Hybrid with OpenMP Less overhead



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WRF Hybrid Scaling

**Compute-Time Performance** (Simulation Time/Compute Time) 400 MPI MPI Tasks Per Node 300 4 MPI Tasks Per Node 2 MPI Tasks Per Node 200 100 Number of Nodes 10 20 30 40 50 60

Samm Elliott. WRF Performance Optimization Targeting Intel Multicore and Manycore Architectures. Apr. 2017

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### RBF-FD Generates weights in scattered node FD formulas

- Total geometric flexibility
- No connexctivities or mappings are required
- Excellent for distributed memory computation
- Easy to achieve high order accuracy
- Local node refinement trivial



Formulation and Benefits of RBF-FD Methods

Scattered nodes:Interpolant: $s(\underline{x}) = \sum_{k=1}^{N} \lambda_k \Psi_k(\underline{x})$ System that determines the expansion coefficients  $\lambda$  $\begin{pmatrix} \Psi_1(\underline{x}_1) & \Psi_2(\underline{x}_1) & \cdots & \Psi_N(\underline{x}_1) \\ \Psi_1(\underline{x}_2) & \Psi_2(\underline{x}_2) & \cdots & \Psi_N(\underline{x}_2) \\ \vdots & \vdots & \vdots & \vdots \\ \Psi_1(\underline{x}_N) & \Psi_2(\underline{x}_N) & \cdots & \Psi_N(\underline{x}_N) \\ \end{pmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_N \end{bmatrix} = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_N \end{bmatrix}$ 

The interpolant is then defined given properties of matrix row swaps:

$$s(\underline{x}) = \sum_{k=1}^{n} \lambda_k \phi(||\underline{x} - \underline{x}_k||)$$

Solve system A = f, of the form:

$$\begin{bmatrix} \phi(\|\underline{x}_1 - \underline{x}_1\|) & \phi(\|\underline{x}_1 - \underline{x}_2\|) & \cdots & \phi(\|\underline{x}_1 - \underline{x}_n\|) \\ \phi(\|\underline{x}_2 - \underline{x}_1\|) & \phi(\|\underline{x}_2 - \underline{x}_2\|) & \cdots & \phi(\|\underline{x}_2 - \underline{x}_n\|) \\ \vdots & \vdots & \ddots & \vdots \\ \phi(\|\underline{x}_n - \underline{x}_1\|) & \phi(\|\underline{x}_n - \underline{x}_2\|) & \cdots & \phi(\|\underline{x}_n - \underline{x}_n\|) \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_n \end{bmatrix} = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{bmatrix}$$





## Summary



### Important Takeaways

- MISO dynamics are internal to monsoon system and hard to predict.
- Effective and efficient air-sea coupling models need to be employed for good simulation results of monsoons.
- Novel algorithms are key to greatest improvements in speed so long as HPC design is made a priority.

### Future Work

- Identify key dynamics of MISO generation for models to address.
- Implement air-sea coupled solvers, including salinity structure to better capture ocean heat content distribution.
- Further investigate DG and RBF-FD Methods for feasibility in NWP and given time, utility in WRF.
- Perform data validation on MISO simulations and compare which models capture MISO signals best.

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Hiking Mullayanagiri peak from east side Western Ghats