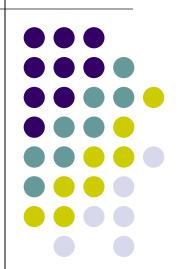
# Impact studies of Data Assimilation using Nudging and 3D VAR for a few atmospheric systems

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#### Introduction

- Modelling / forecasting of Tropical cyclones requires accurate initial/ boundary conditions to precisely forecast the intensity and movement of the developing storm and its landfall point
- 3-DVAR and Newtonian Nudging are widely used assimilation techniques to incorporate observations in to a dynamical model
- However they require to be evaluated for different sources of data, observational error coefficients (3dvar), period and strength of nudging to assess their potential in data assimilation
- Satellite Scatterometer wind observations, temperature/ humidity profiles and conventional land based observations are potential sources of data for prediction of various atmospheric systems.

### **Objectives**

- To study the impact of vector wind observations from QuickSCAT (Quick Scatterometer) and ASCAT (Advanced Scatterometer) at 25 km resolution using Four Dimensional Data Assimilation (FDDA) analysis nudging technique in WRF ARW nested mesoscale model on simulation of wind field (for dispersion) and Tropical Cyclones.
- To assess the effect of nudging on the intensity and movement with respect to period of nudging and strength of nudging
- To study the effect of assimilating different data sets (QucikSCaT/SSMI winds, conventional observations) on cyclone simulations using 3D-Var

## **Newtonian Relaxation or Nudging**

The widely used method for data assimilation is *Newtonian relaxation* of *nudging which* consists of adding an empirical term to the prognostic equations that nudges the solution towards the observations



$$\frac{\partial \alpha}{\partial t} = F + G_{\alpha}W(x, y, \sigma, t) \cdot (\alpha_0 - \alpha),$$

Where  $\mathbf{F}$  = All model's Physical Processes

 $G_{\alpha}$  = Positive relaxation term which determines relative weight of the relaxation term for a given parameter

**W** = Four dimensional weighing Function

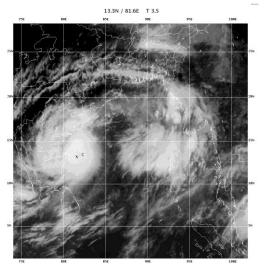
 $\alpha_{o}$  = Gridded field of  $\alpha$  obtained from objective Analysis of Observations

#### **Questions**

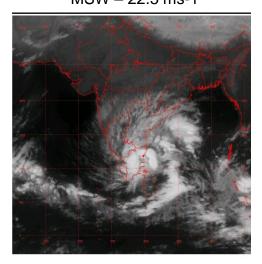
- •Which data are useful? i.e., Impact of different kinds of observations
- •How much period is sufficient to obtain better initial condition?
- •What strength of nudging to be applied?

#### A few cyclonic systems in the present study....

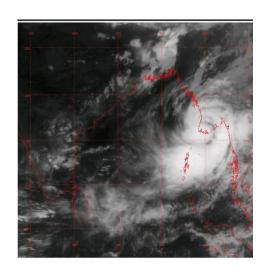
LAILA (18 –21 May 2010)



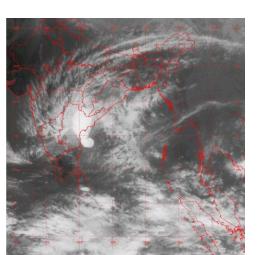
NISHA (24-28 Nov 2008) CSLP – 996 h Pa MSW – 22.5 ms-1



NARGIS (27Apr-2May 2008)



KAIMUKH (15-15 Nov 2008)



SIDR (11-16 Nov 2007 CSLP – 944 hPa; Max winds – 115 knots



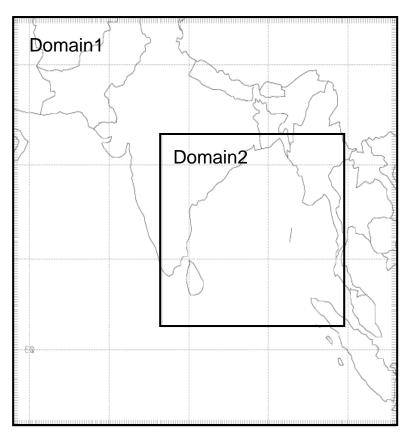
Land fall

Laila – Near Machilipatnam Nargis – Mynmar Sidr – Bangladesh Nisha – Near Chennai Kaimukh – Near Machilipatnam





Dynamics	Primitive equation, non-hydrostatic		
Vertical resolution	28 levels		
Domains	Domain1	Domain2	
Horizontal resolution	27 km	9km	
Grid points	173 x 162	249 x 239	
Domains	58.02-105.97 E; -8.26 - 34.31 N;	77.9-99.40 E; 4.49-23.60 N	
Radiation	Dudhia scheme for shortwave and RRTM for long wave processes		
SST	NCEP FNL analysis data		
Convection	Kainfritch Scheme		
Explicit moisture	Lin et al		
PBL turbulence	YSU PBL		
Surface processes	NOAH LSM scheme		



#### **Simulations**

- Model integrated for 72 hours using the NCEP GFS analysis as first guess for initial/ boundary conditions
- Control run uses NCEP GFS data alone for IC/BC, in assimilation runs the first guess is modified with various observations
- In nudging experiments observations are combined with background first guess fields using objective interpolation / SCM methods (viz., Cressman).
- Nudging experiments are conducted with respect to period of nudging (6 hour, 12 hour, 18 hour, 24 hour) during pre-forecast period
- Strength of nudging for winds (1.0e-4, 2.0e-4, 3.0e-4, 4.0e-4, 5.0e-4)

#### Sources of Data

- Conventional (surface, upper air Radiosonde)
- Satellite (QSCAT, SSM/I, ASCAT winds, MODIS temperature, humidity profiles)
- ISRO Automated Weather Stations



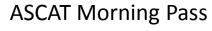
- QSCAT is NASA's scatterometer operates with frequency of 13.4GHZ and provides wind vector with resolution of 25KM at 10m level
- QSCAT has a swath width of 18000KM and measures wind speed with in the range of 3 to 20m/s
- The products of QSCAT are wind speed, wind direction at 10mlevel total columnar water vapor and rain rate estimates

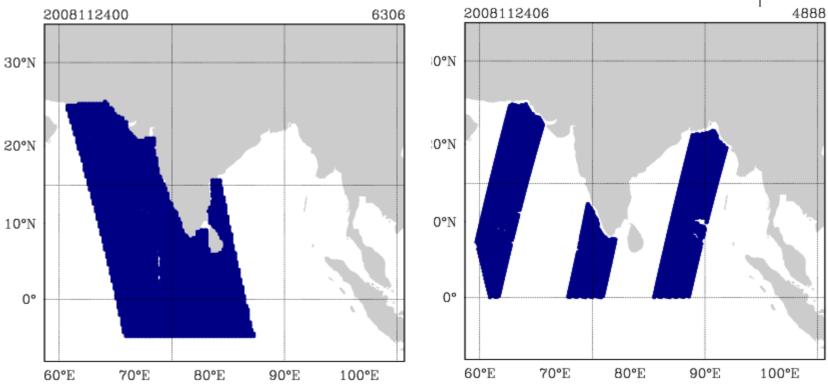
- Advanced scaterometer is EURUMETSAT's scaterometer operates with frequency of 13.4GHZ and provides wind vector with resolution of 25KM at 10m level
- ASCAT has a swath width of 9000KM and measures wind speed with in the range of 3 to 25m/s
- The products of ASCAT are wind speed, wind direction at 10mlevel and soil moisture estimates



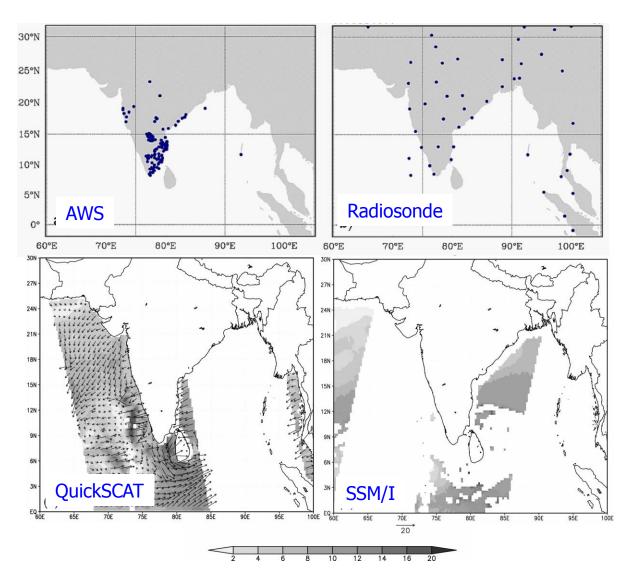


## 





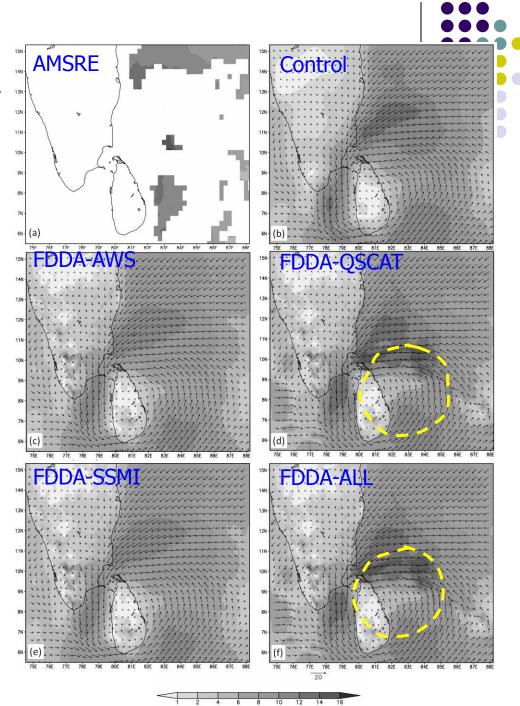
### Experiments with assimilation of different observations



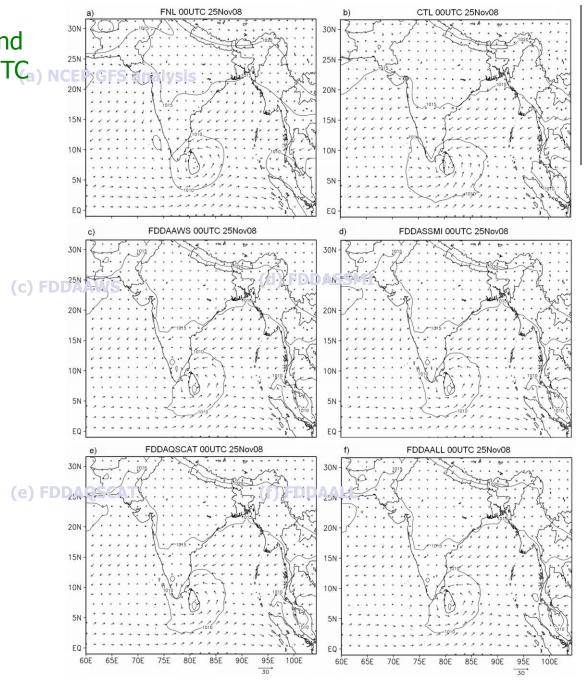


### Cyclone - NISHA

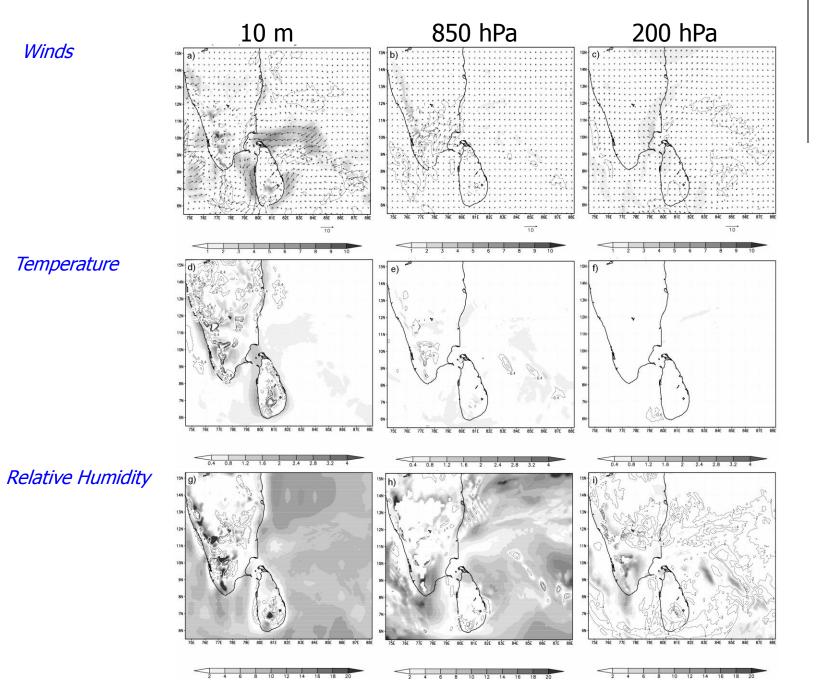
Comparison of 10 m winds from different analysis at 00 UTC 25 Nov



Seal level pressure and surface winds at 00UTC 25 Nov 2008



#### Difference in initial fields between FDDA and CONTROL 00 UTC 25 Nov 2008 • •



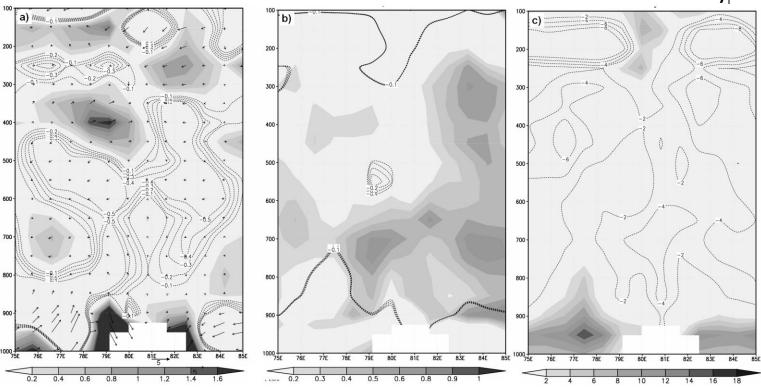
## Vertical section of difference in initial fields FDDA and CONTROL runs for 25 Nov 2008, 00 UTC at 7 $^{\circ}$ N

**Analysis increments** 

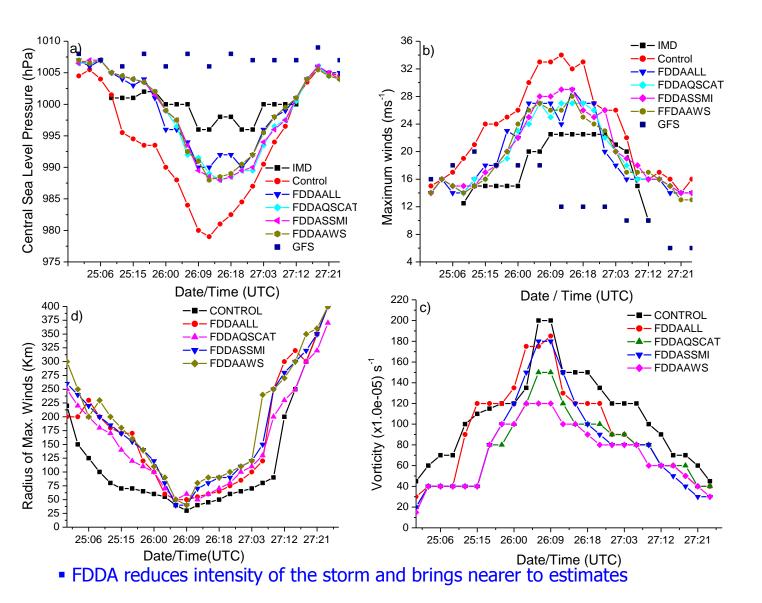
Horizontal winds

Potential temperature

Relative humidity



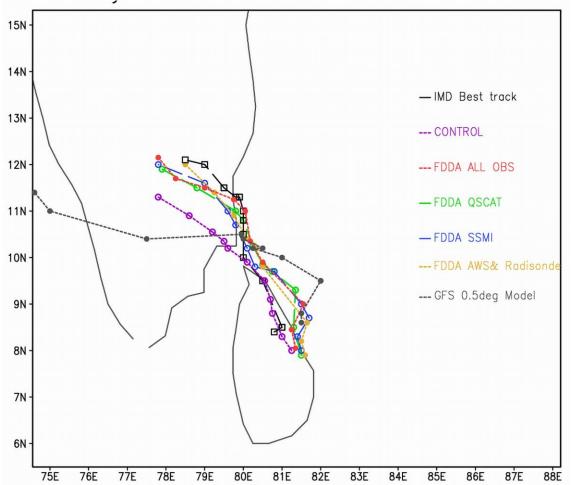




- FDDAQSCAT, FDDASSMI almost similarly perform except during peak development stages where QSCAT shows large impact with substantial reduction in the intensity
- FDDAQSCAT, FDDAALL gives the best simulations for CSLP, winds



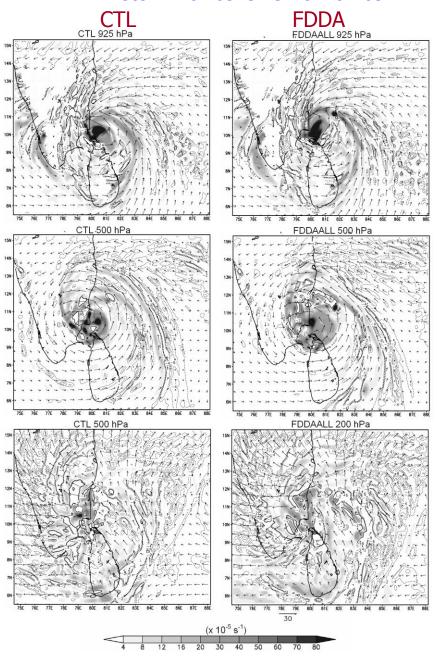
#### Simulated track estimates



- FDDA QSCAT runs produce the minimum errors in the development phase
- FDDA AWS runs produce minimum errors during landfall phase
- FDDAALL (with all observations) produce the minimum track errors through out simulation



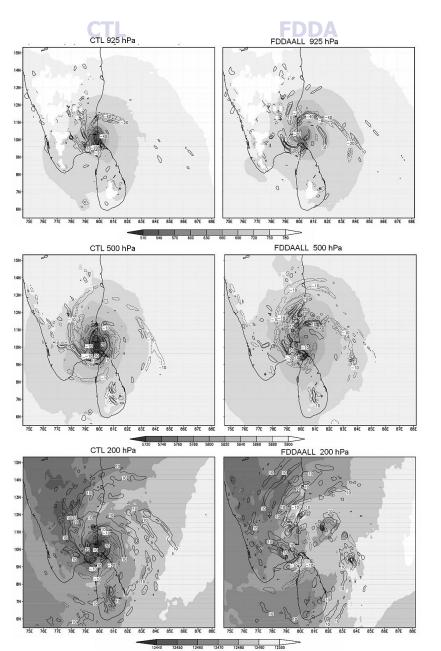
## Wind field and Vorticity (x10-5 s-1) associated with the simulated storm for 09 UTC 26 Nov 09.



Assimilation of all observations shows improvement in simulations over control run with respect to wind field and vorticity and movement of the storm



Geopotential (m) and convergence/divergence (x10-5 s-1) associated with the simulated storm for 09 UTC 26 Nov 09.

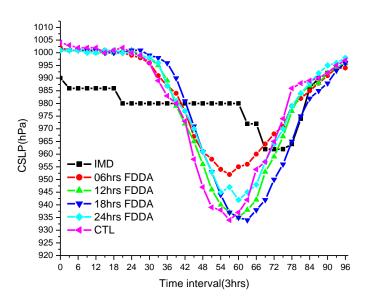


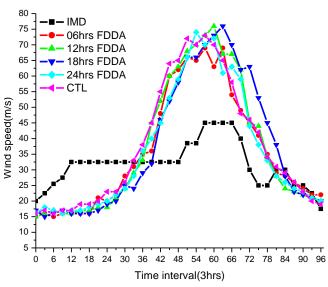


## Simulated Rainfall pattern in CTL, FDDA runs along with TRMM data CTL 00UTC26-27UTC NOV08 FDDAALL 00UTC26-27UTC NOV08 TRMM 00UTC26-27UTC NOV08 Control TRMM 00UTC27-28UTC NOV08 CTL 00UTC27-28UTC NOV08 FDDAALL 00UTC27-28UTC NOV08 **FDDAALL** 100 120 140 160 180 200 220 240 260

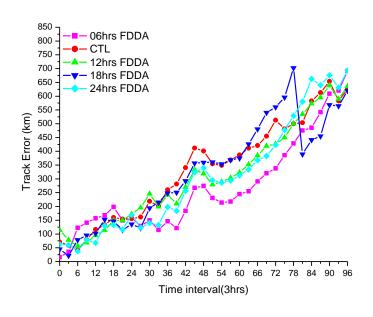
The distribution and pattern of rainfall is relatively better simulated with FDDAALL (with inclusion of all observations)

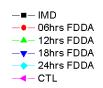
## **Experiments with different Nudging Periods**





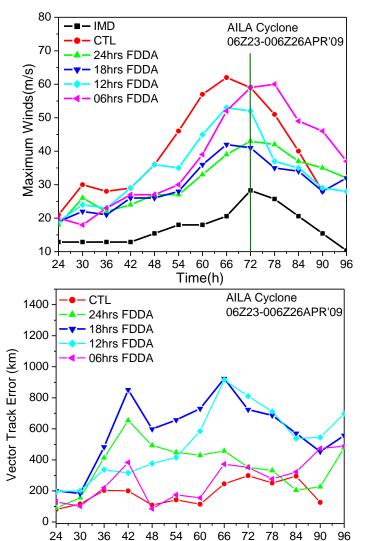
- •FDDA reduced the intensity of the simulated storm and the maximum effect is found with 6h nudging
- •Error in CSLP is minimum with 6hr/ 24 hr nudging and maximum with 12/18 h
- •Error in winds is minimum with 6r nudging and increases with period of nudging further
- •Errors in track positions are improved after 24h simulation and minimum with 6h nudging



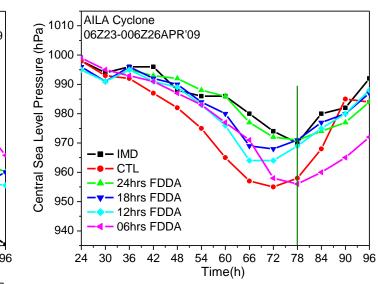


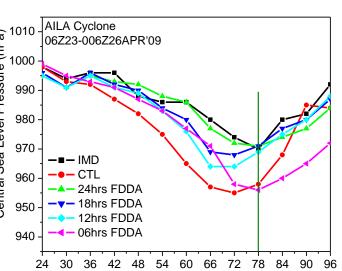
### Experiments with different Nudging Periods

AILA Cyclone 06Z23 - 006 Z 6 Apr 2009 Conventional data alone used in Assimilation



Time(h)

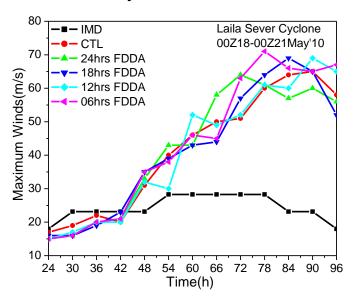


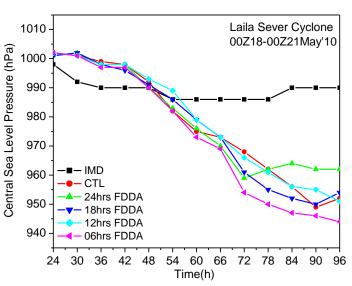


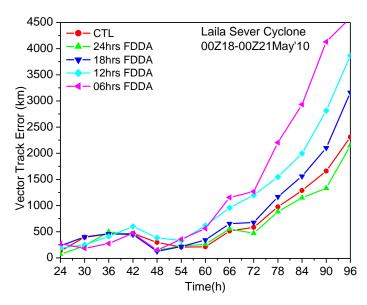
- •FDDA reduced the intensity of the simulated storm and the maximum effect is found with 18/24 h nudging
- •Error in CSLP is minimum with 18/24 h nudging
- •Error in winds is minimum with 18/24 h nudging
- •Errors in track positions are minimized with 6h /24h of nudging



#### LAILA Cyclone 00Z18 - 00Z21 May 2010



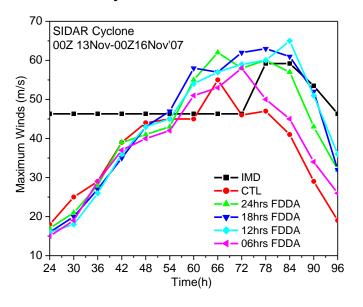


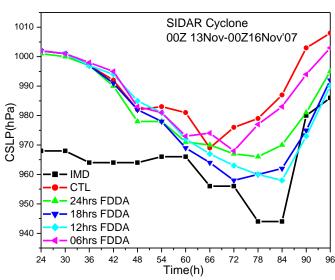


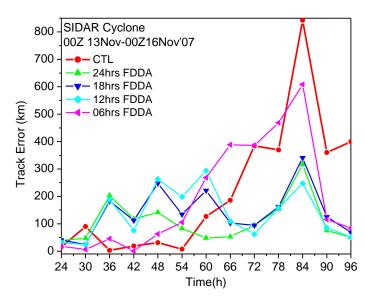
- FDDA marginally reduced the intensity of the simulated storm and the maximum effect is found with 24 h nudging
- Errors in track positions are minimized with 18/ 24h of nudging



#### SIDAR Cyclone 00Z13 - 00Z16 Nov 2007



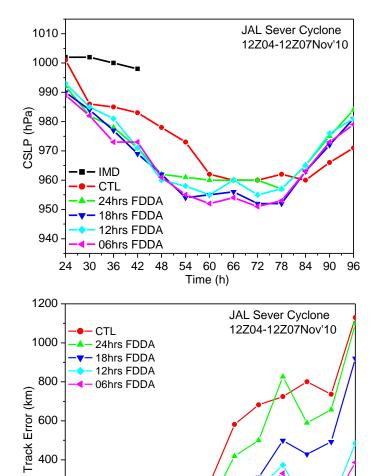




- Nearest simulated intensity of storm is give by FDDA 18/24 h nudging
- Errors in track positions are minimized with 18 /24h of nudging



#### JAL Cyclone 12Z04 -12Z07 Nov 2010



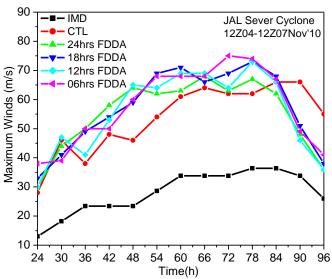
200

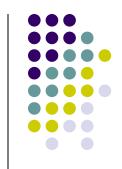
30

36 42 48

54 60 66 72 78 84 90 96

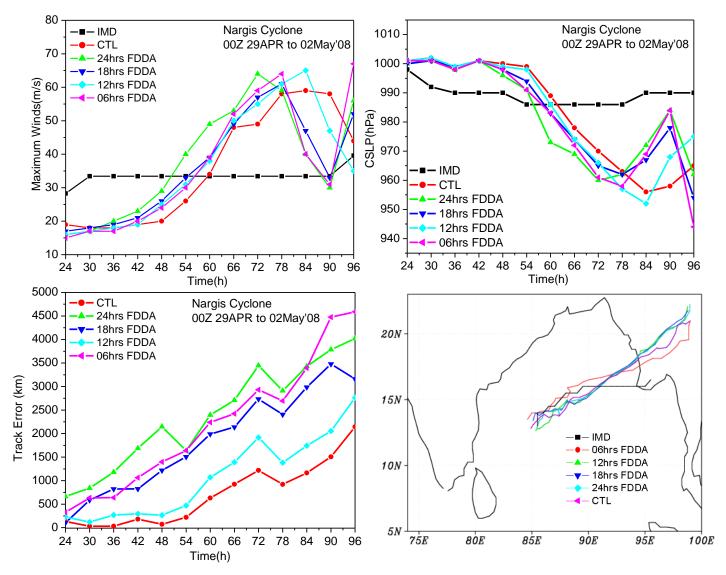
Time(h)





- No significant effect found on intensity of storm with FDDA nudging
- •Errors in track positions are minimized with 6h /12h of nudging

#### NARGIS Cyclone 00Z29 -00Z02 May 2008



- No significant effect found on intensity of storm with FDDA nudging
- Errors in track positions are minimized with 6h /18h of nudging



#### Nisha (24-28 Nov 2008) 50 — IMD 1010 45 06hrs FDDA - 12hrs FDDA 7— 18hrs FDDA 1005 40 24hrs FDDA CTL 35 Wind speed (m/s) 1000 CSLP(hPa) 995 990 - 06hrs FDDA -A- 12hrs FDDA 15 -v- 18hrs FDDA 985 24hrs FDDA 10 CTL 980 60 66 72 78 30 36 42 54 72 170 1N — IMD 160 --- CTL 06hrs FDDA 150 - 06hrs FDDA 12hrs FDDA 140 12hrs FDDA -▼- 18hrs FDDA 130 24hrs FDDA 3*N* 24hrs FDDA 120 CTL Track Error (km) 100 90 ЭN 60 3*N* 50 40 30 20 3N

•The maximum effect is found with 12h/24 h nudging

33 36 39 42 45 48 51 54 57 60 63 66 69 72 75 78 81 84

Time interval(3hrs)

- •Error in CSLP is minimum with 12hr nudging and maximum with 18 h
- •Error in winds is minimum with 12hr nudging and increases with period of nudging further

74E

76E

78E

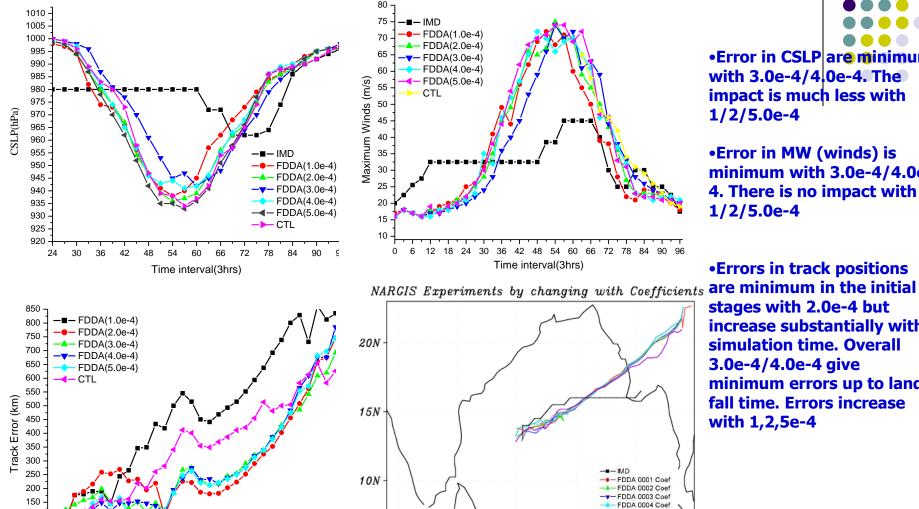
82E

84E

80E

•Errors in track positions are minimum with 12h/24h nudging throughout simulation

#### Experiments considering different strengths of nudging



5Λ

75E

8ÒE

85E

9ÒE

36 42 48 54 60 66 72 78 84 90 96

Time interval(3hrs)

12 18 24 30

•Error in CSLP are minimum with 3.0e-4/4.0e-4. The impact is much less with 1/2/5.0e-4

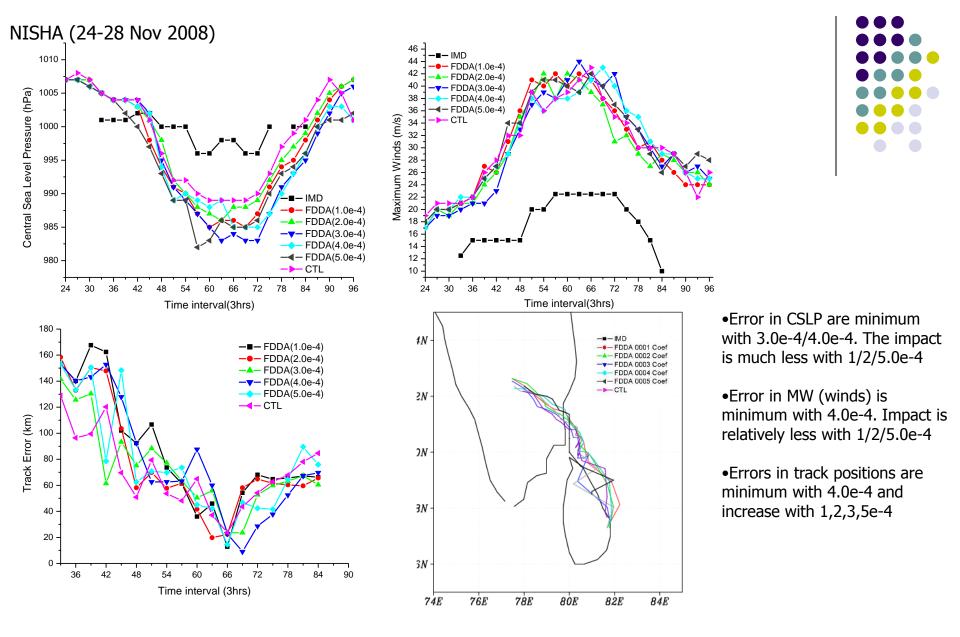
•Error in MW (winds) is minimum with 3.0e-4/4.0e-4. There is no impact with 1/2/5.0e-4

 Errors in track positions stages with 2.0e-4 but increase substantially with simulation time. Overall 3.0e-4/4.0e-4 give minimum errors up to land fall time. Errors increase with 1,2,5e-4

→ FDDA 0005 Coef --- CTL

95E

100E



#### **Main Results**

- ❖The vlaue 3.oe-4/4.0e-4 is found to be optimum for nudging of wind observations
- ❖A pre-forecast period of 12-18 h nudging is sufficient for optimum forecasts

#### Application of Nudging to simulation of wind field and dispersion

A few cases of wind flow were studied with assimilation of QuickSCAT and conventional observations in NCAR/MM5 model

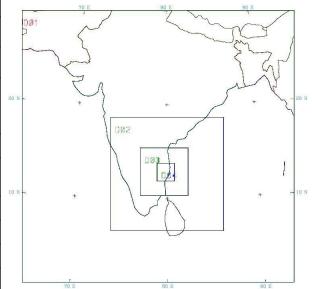


#### Observations 1. QSCAT

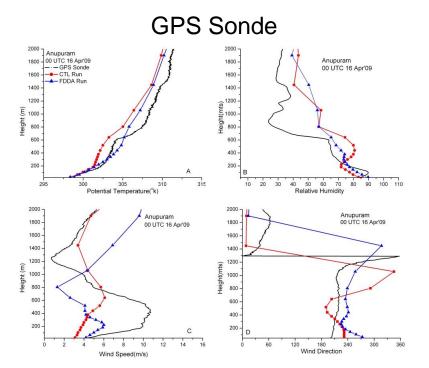
- 2. Surface AWS
- 3. Upper Air Radiosonde

Dynamics	Primitive equation, non-hydrostatic			
Vertical resolution	32 levels			
Domains	Domain1	Domain2	Domain3	Domain4
Horizontal resolution	36 km	12 km	4 km	1.33 km
Grid points	90 x 90	112 x 112	142 x 142	157 x 157
Domains of integration	62.61-95.37 E; 0.80-30.15 N	73.69-86.40 E; 6.49-18.60 N	77.04-82.31 E; 10.22-15.33 N	78.88-80.80 E; 11.81-13.69 N
Radiation	Dudhia scheme for shortwave and RRTM for longwave processes			
Sea surface temperature	NCEP FNL analysis data			
Convection	Grell scheme on the outer grids domain1, domain2			
Explicit moisture	Simple ice (SI) scheme			
PBL turbulence	Hong-Pan (MRF) PBL			
Surface processes	5 layer soil thermal diffusion scheme			

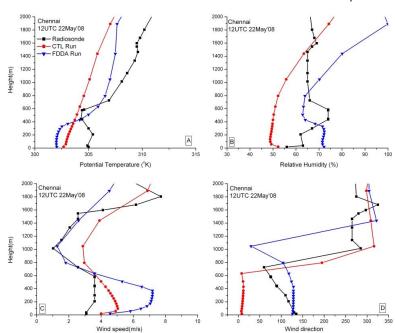
## New analysis prepared using Little\_R objective analysis



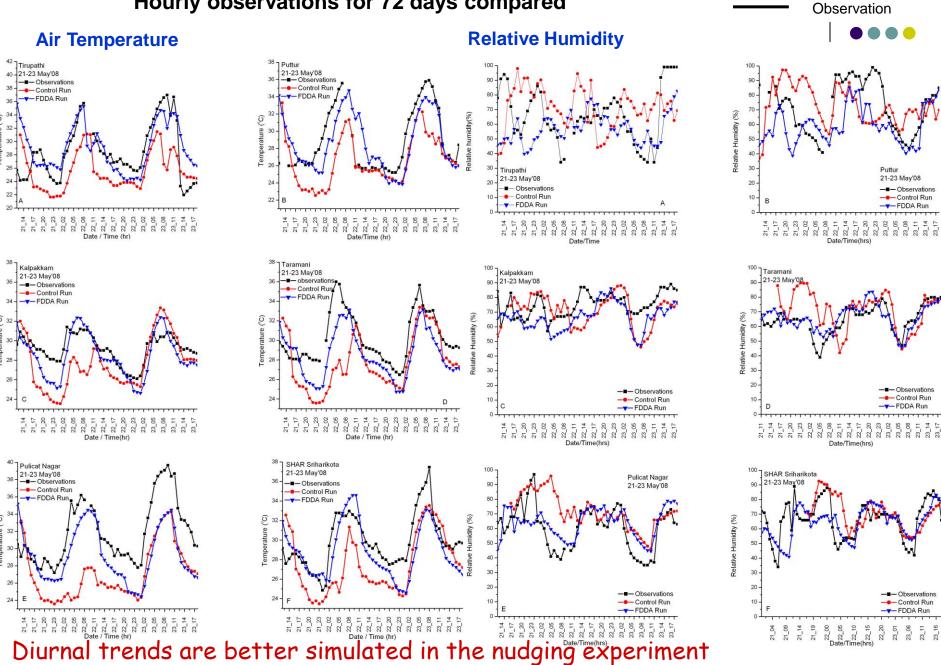
## Comparison of model profiles with observations



#### RS/RW Chennai

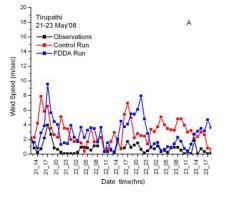


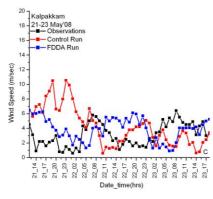
Model results for 25 stations analysed Hourly observations for 72 days compared

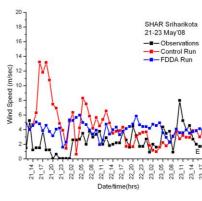


No assimilation Assimilation

#### Wind speed







#### Wind direction



8 9 9 2 2 2 Date/Time(hrs)

22\_20

Taramani

Wind Direction

100

21-23 May'08

400 TSHAR Sriharikota

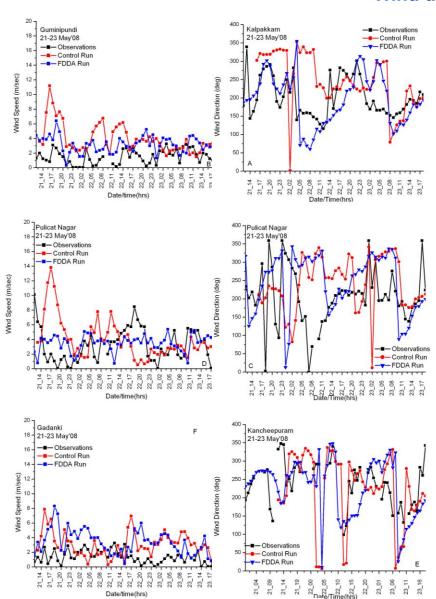
21-23 May'08

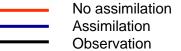
─■─ Observations
── Control Run

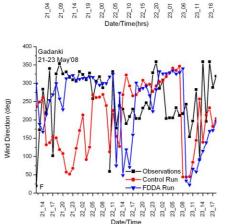
-■- Observations

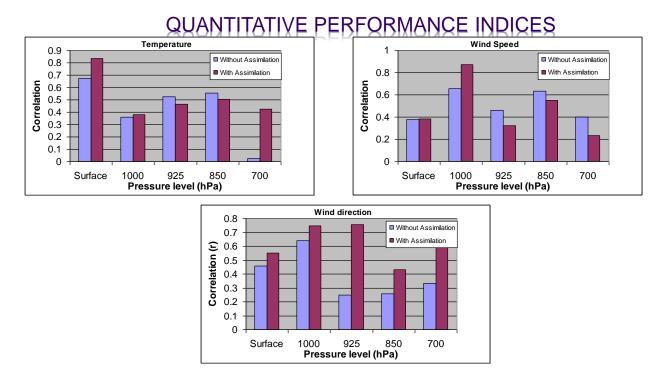
Control Run

▼- FDDA Run



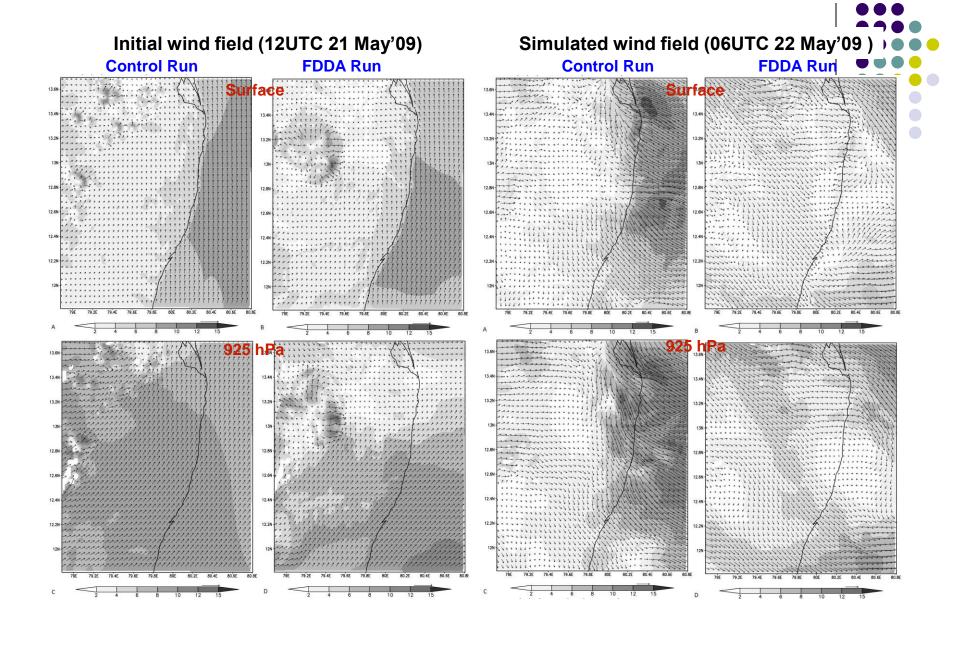




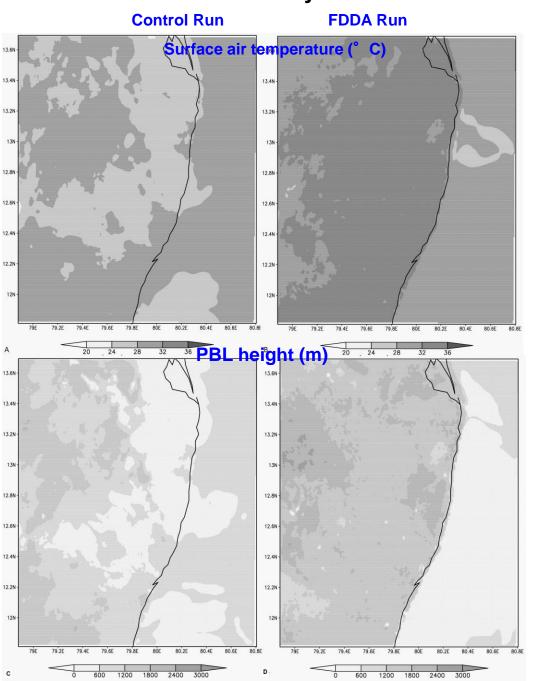


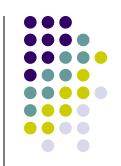
- •There were differences in the flow field, temperature, humidity and mixing height in the two cases which differently simulated the plume dispersion.
- •Assimilation of observations led to improved statistics (r,s,mb,me,rmse etc) for various parameters of interest.

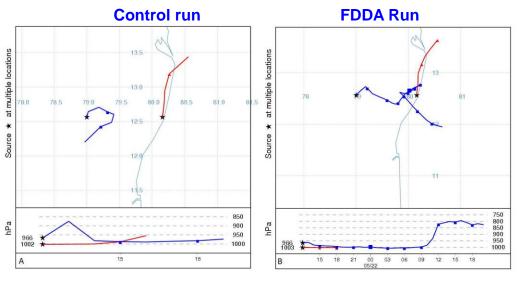




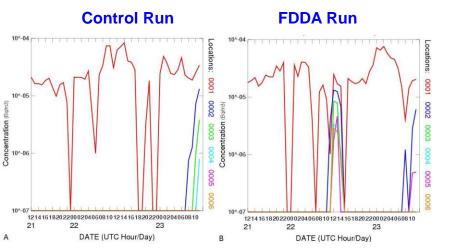
#### 06UTC 22 May'09.







#### Time sereis of Receptor concentrations across coast

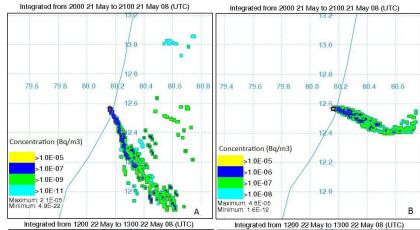


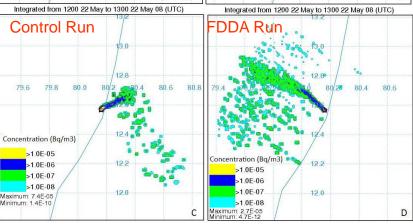
• Winds from FDDA experiment direct the plume onshore while in control experiment the plume is offshore 14-17 h LT 22 May'08

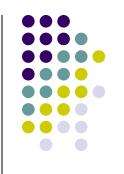


#### **Control Run**

#### **FDDA Run**



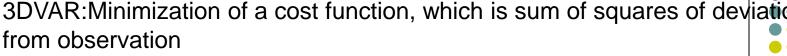




Assimilation of dense AWS observations with QsCAT winds in the FDDA case seems to produce large impact on the performance of the MM5 model for the parameters of diurnal wind field, temperature, mixing height in the domain of interest.

Comparisons with observations indicated better statistics with the FDDA fields than in the control run. The simulated plume, its concentration distribution varied in the two cases and shows realistic results from the MM5FDDA fields.

### 3D-VAR Method – A case study for a cyclone on southeast coast with WR



The cost function (J) is defined as

$$J(\mathbf{x}) = \frac{1}{2} (\mathbf{x} - \mathbf{x}^b)^{\mathsf{T}} \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}^b) + \frac{1}{2} (H(\mathbf{x}) - \mathbf{y}^o)^{\mathsf{T}} \mathbf{R}^{-1} (H(\mathbf{x}) - \mathbf{y}^o)$$
Background term
Observation term

B = background error covariance matrix

R = observation and representativeness error covariance matrix

 $H = \text{nonlinear observation operator (model space } \rightarrow \text{ observation space)}$ 

• The minimization requires an estimation of the gradient of the cost function:

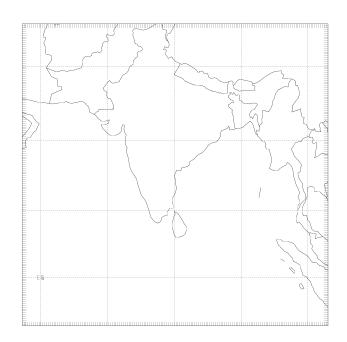
$$\nabla J(\mathbf{x}) = \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}^b) + \mathbf{H}^{\mathrm{T}}\mathbf{R}^{-1}(H(\mathbf{x}) - \mathbf{y}^o)$$

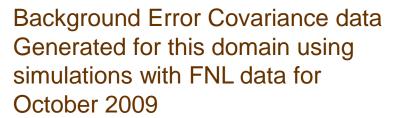
$$(X - X_b)$$
 = Analysis increments

(
$$H(X_b)$$
-  $y^o$ ) = Innovation vector

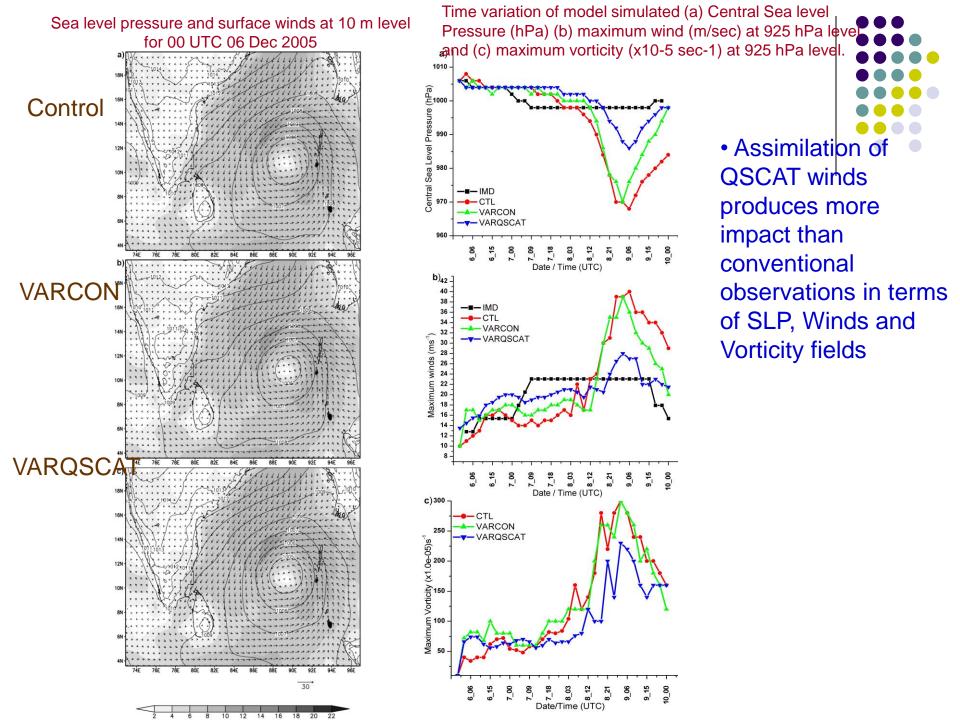
### Case study of Fanoos cyclone 4- 10 Dec 2006

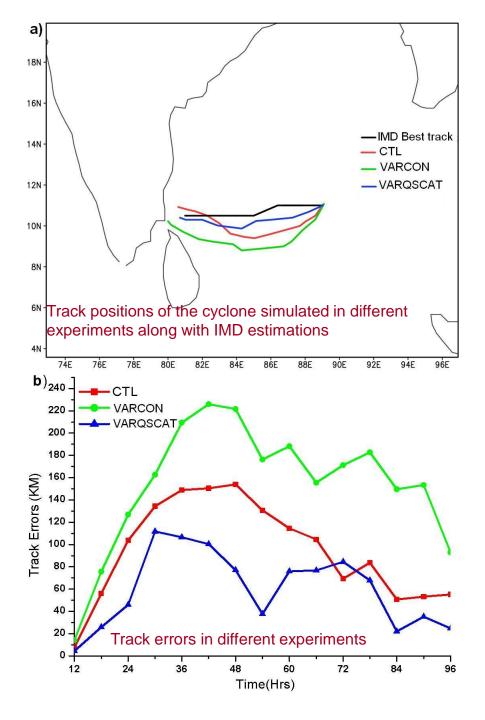
Dynamics	Primitive equation, non- hydrostatic
Vertical resolution	28 vertical levels
Horizontal resolution	27 km
Domains of integration	74.50 E – 82.50 E (162X164) 9.00 N – 16.74 N
Radiation	Dudhia scheme for short wave radiation Rapid Radiation Transfer Model (RRTM) for long wave radiation
Explicit moisture	Lin microphysics (Lin et al., 1983)
Surface processes	5-layer soil thermal diffusion scheme
Planetary boundary layer	Yonsai University (YSU) PBL
Cumulus convection	Kain-Fritisch (KF-Eta)
Sea-surface temperature	Real SST





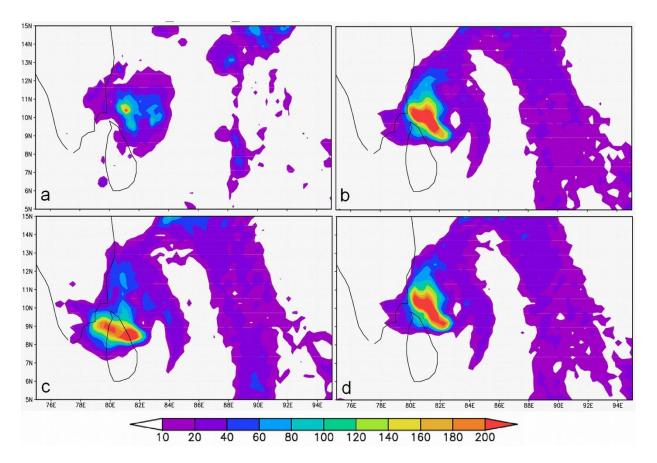








- Assimilation of conventional data gives large errors with simulated cyclone moving to the south of observed storm
- Assimilation of QSCAT winds reduce the track errors considerably



Accumulated 24 h rainfall (mm/day) ending at 00 UTC 10 Dec 2005 from a) TRMM data b) CONTROL c) VARCON and d) VARQSCAT L experiments respectively



#### **Conclusions**

- FDDA nudging experiments using ARW with scatterometer wind observations indicates nudging reduced the overintensification of storm in the control run as well as reduced the track errors.
- A period of 12h/18 h nudging during the pre-forecast stage provides maximum impact on simulations.
- Nudging effect is maximum with coefficient of nudging for winds 3.0e-4 4.0e-4 which corresponds to a time scale of one hour.
- Simulations with assimilation of several individual data sets (AWS, Qscat winds, SSM/I winds, Alldata) clearly show that the movement of the storm while approaching the coast is substantially improved using all observations.
- Vector track errors are less in the FDDAQSCAT than in FDDASSMI and FDDAAWS cases upto the landfall time of the simulated storm
- The FDDAQSCAT has better produced the intensity of the cyclone (interms
  of maximum winds) than the FDDASSMI which may be because of better
  representation of wind field from QSCAT vector wind observations
- Satellite winds give higher impact using 3DVAR for cyclone simulation than the conventional observations



# -Thank-You