

Applications of Satellite Data to Tropical Cyclone Forecasting

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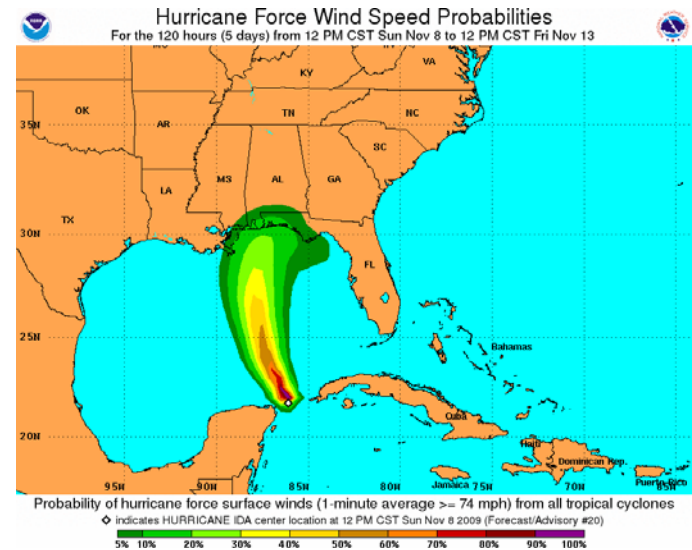
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Theme IV- Severe Weather and Hazards



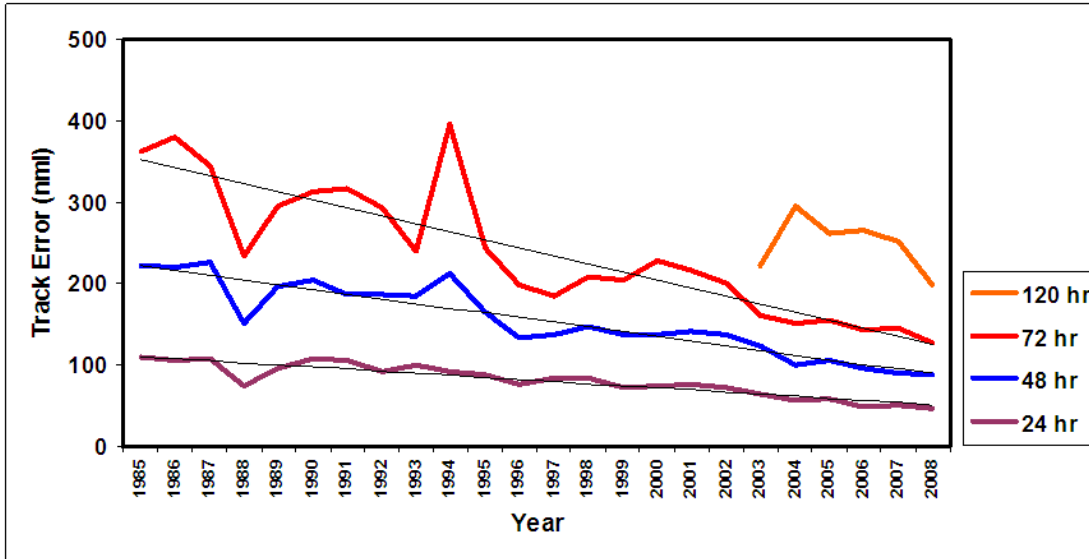
National Hurricane Center Tropical Cyclone Forecast Parameters

- Track
 - tropical cyclone positions 0 to 120 hr
- Intensity
 - maximum wind 0 to 120 hr
- Wind Structure
 - Radii of 34, 50 kt winds 0 to 72 hr
 - Radii of 64 kt winds 0 to 36 hr
- Watches
 - Hurricane conditions possible in next ~36 h
- Warning
 - Hurricane conditions likely in next ~24 h
- Probabilities of 34, 50 and 64 kt winds 0 to 120 hr
- General guidance on rainfall and storm surge
- Probability of TC formation in the next 48 h
- *All forecast products updated every 6 hours*

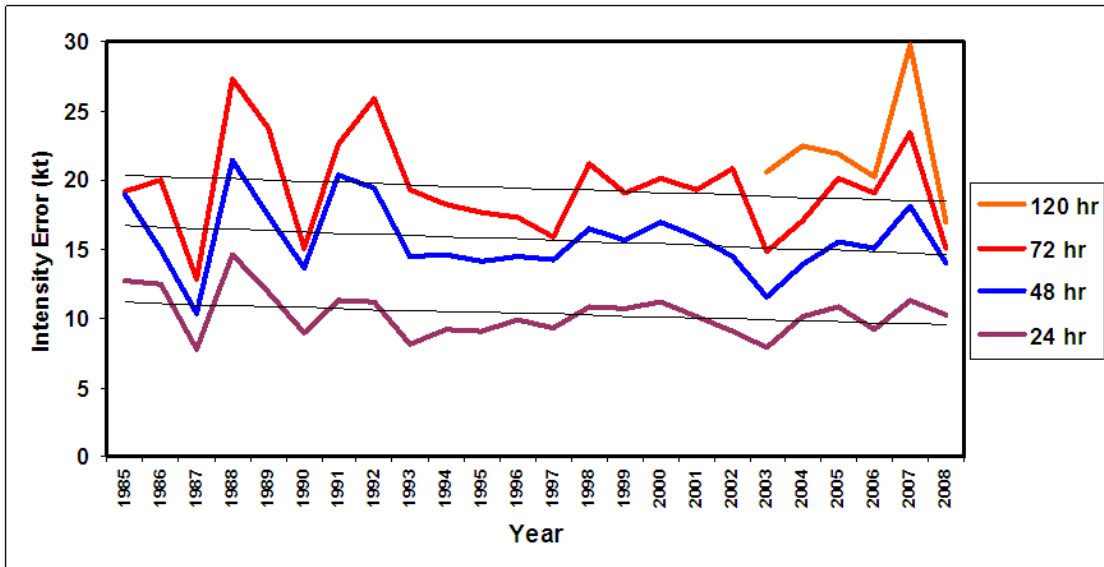


Hurricane wind probability product for Hurricane Ida 8 Nov 2009, 18 UTC

Mean Absolute Error of NHC Official Atlantic Track and Intensity Errors 1985-2008



**48 hr Track Improvements
~3.7% per year**



**48 hr Intensity Improvements
~0.6% per year**

Models for Operational Tropical Cyclone Track and Intensity Forecasting

- Track
 - Global models: **NCEP-GFS**, **ECMWF**, NOGAPS, UKMet
 - Regional models: **GFDL**, **HWRF**
 - Both include coupled ocean model
 - Simple climatology and persistence (CLIPER) for skill baseline
 - Simple trajectory models
- Intensity
 - Regional models: GFDL, HWRF
 - Statistical-dynamical models: **SHIPS**, **LGEM**
 - Rapid Intensity Index: Discriminant analysis technique
 - Simple climatology and persistence (SHIFOR) for skill baseline
- **Models in red were most accurate over the last 5 years**

Satellite Data Utilization in Hurricane Forecasting

- Identify cloud clusters that may form TCs
- Center location and intensity estimates
 - Vis/IR Dvorak techniques, microwave methods
- Synoptic feature identification
 - Feature track winds, WV imagery
- Wind structure analysis
 - Scatterometer, passive microwave
- Sea surface temperature products
- Assimilation into numerical models
 - Atmosphere and ocean
- Predictors in statistical intensity forecast models
 - Oceanic heat content from satellite altimetry
 - GOES data for convective analysis
 - Microwave imagery
 - Lightning data on GOES-R using proxy ground based systems



Satellite data used in NCEP's operational data assimilation systems

- HIRS sounder radiances
- AMSU-A sounder radiances
- AMSU-B sounder radiances
- AIRS radiances
- IASI radiances
- GOES sounder radiances
- GOES, Meteosat, GMS winds
- GOES precipitation rate
- SSM/I precipitation rates
- TRMM precipitation rates
- SSM/I ocean surface wind speeds
- ERS-2 ocean surface wind vectors
- ASCAT ocean surface wind vectors
- QuikSCAT ocean surface wind vectors*
- AVHRR SST
- AVHRR vegetation fraction
- AVHRR surface type
- Multi-satellite snow cover
- Multi-satellite sea ice
- SBUV/2 ozone profile and total ozone
- AIRS
- MODIS Winds
- Altimeter sea level observations (ocean data assimilation and wave data assimilation system)

Satellite data also used to help estimate initial storm position, motion, intensity and wind structure for “TC Vitals” file used to initialize regional models

Tropical Cyclone Research Projects at NESDIS/RAMMB and CIRA

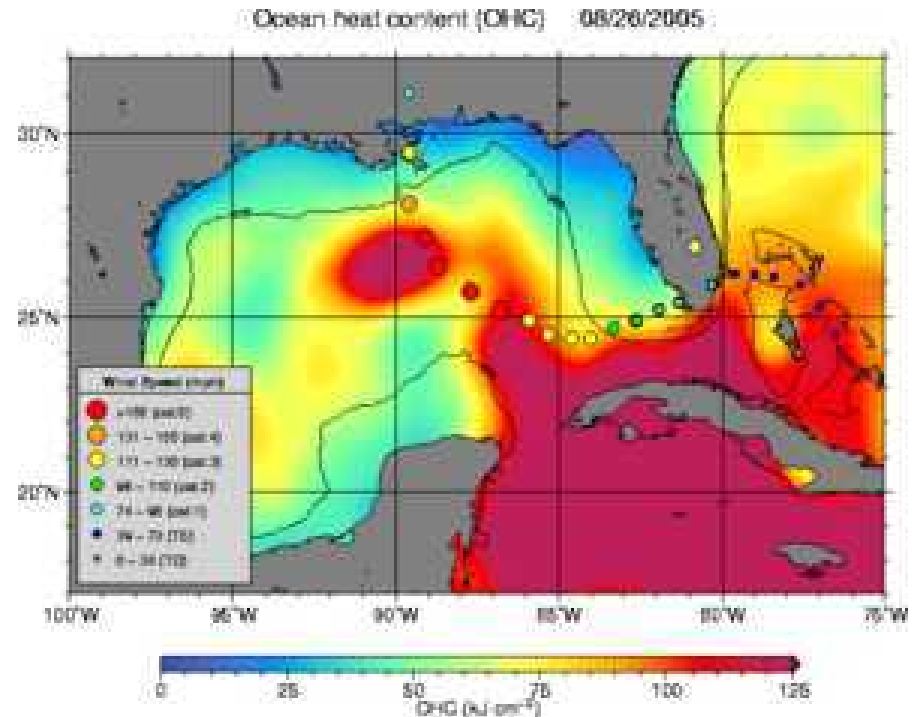
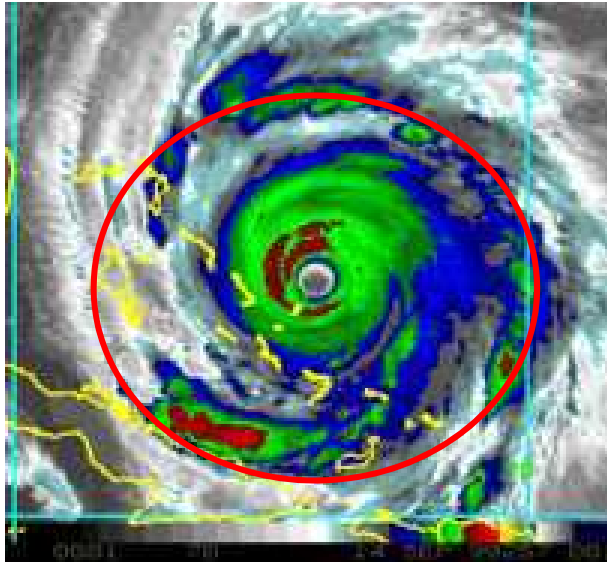
- Improvements to operational statistical hurricane intensity models using satellite data
 - Better use of GOES data
 - Oceanic Heat Content from satellite altimetry
 - Inclusion of passive microwave imagery
 - Advanced statistical algorithms
 - GOES-R studies using proxy data
 - SEVERI for ABI and ground-based lightning for GLM
- Tropical cyclone genesis prediction
 - Combining model, GOES, climatology data in a discriminant analysis
- Multi-platform estimation of tropical cyclone intensity and structure
- Advanced satellite data assimilation for tropical cyclone models
- Development and improvements of NHC's wind speed probability model

Statistical/Dynamical Intensity Models

SHIPS (Statistical Hurricane Intensity Prediction Scheme)

- Multiple regression model
- Predictors from climatology, persistence, atmosphere and ocean
 - Atmospheric predictors from GFS forecast fields
 - SST from Reynolds weekly fields along forecast track
 - Predictors from satellite data
 - Oceanic heat content from altimetry
 - GOES IR window channel brightness temperatures
- Decay SHIPS
 - Climatological wind decay rate over land

Satellite Predictors added to SHIPS in 2003



1. GOES cold IR pixel count
2. GOES IR T_b standard deviation

3. Oceanic heat content from satellite altimetry (TPC/UM algorithm)

Cold IR, symmetric IR, high OHC favor intensification

The Logistic Growth Equation Model (LGEM)

- Applies a simple differential equation to constrain the max winds between zero and the maximum potential intensity
 - Based on analogy with population growth modeling
- Intensity growth rate predicted using SHIPS model input
- More responsive than SHIPS to time changes of predictors such as vertical shear
- More sensitive track errors
- More difficult to include persistence

Logistic Growth Equation (LGE) Model

$$dV/dt = \kappa V - \beta (V/V_{mpi})^n V$$

(A) (B)

Term A: Growth term, related to shear, structure, etc

Term B: Upper limit on growth as storm approaches its maximum potential intensity (V_{mpi})

LGEM Parameters:

- $\kappa(t)$ Growth rate (from SHIPS parameters)
- β MPI relaxation rate (constant)
- $V_{mpi}(t)$ MPI (from SST)
- n “Steepness” parameter (constant)

LGE replaced by Kaplan and DeMaria inland wind decay model over land

Analytic LGE Solutions for Constant κ and

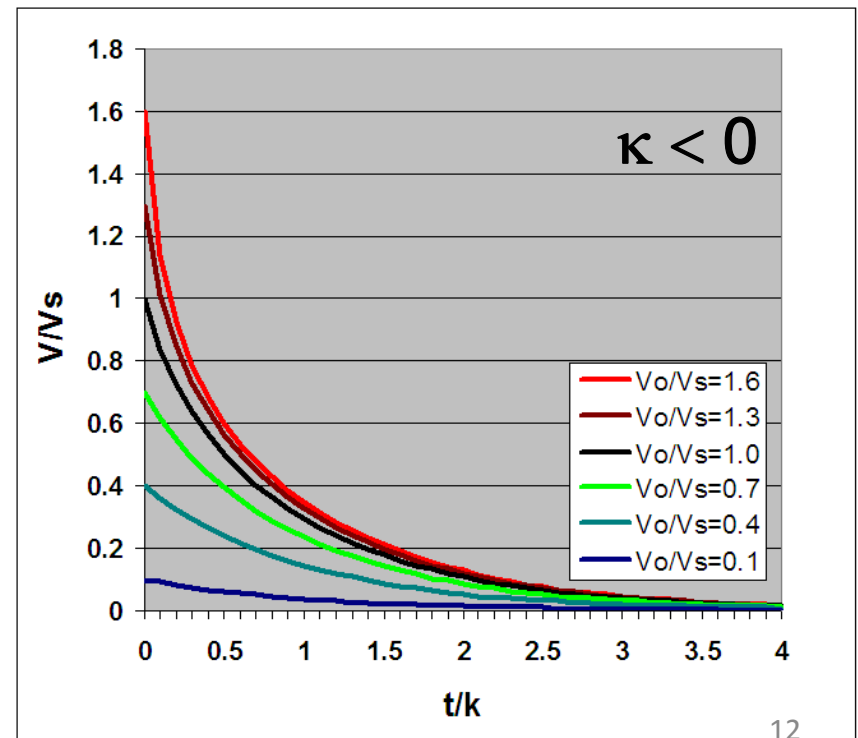
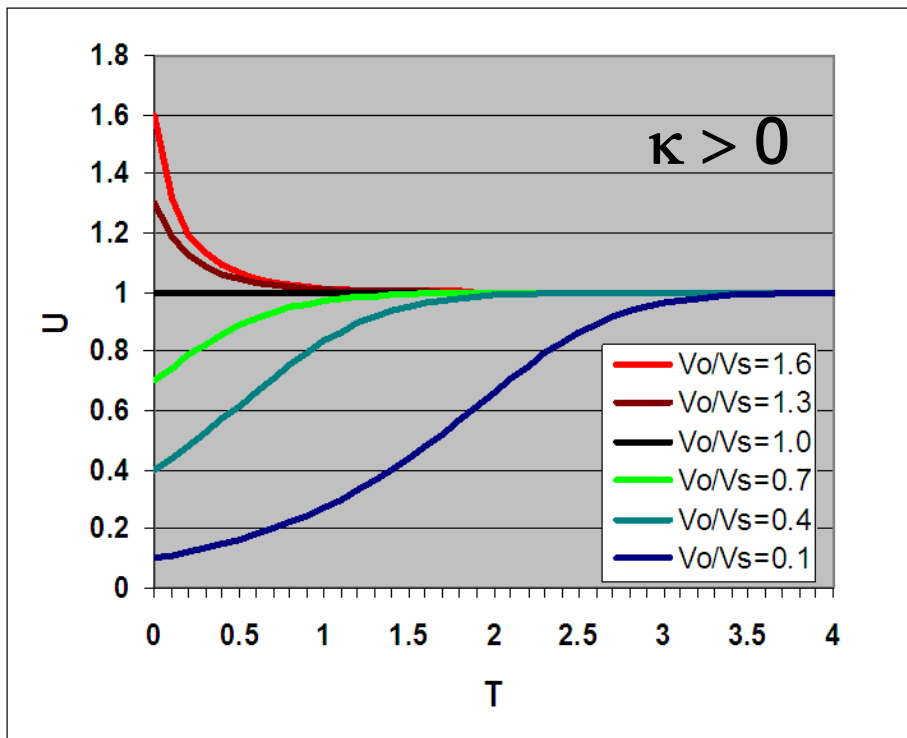
$$V_{\text{mpi}}$$

$$V_s = \text{Steady State } V = V_{\text{mpi}}(\kappa/\beta)^{1/n}$$

$$\text{Let } U = V/V_s \text{ and } T = \kappa t$$

$$dU/dT = U(1-U^n)$$

$$U(t) = U_o \{e^{nT} / [1 + (e^{nT} - 1)(U_o)^n]\}^{1/n}$$



Application of Steady State Solution

$$V_s = \text{Steady State} = V_{\text{mpi}} (\kappa/\beta)^{1/n}$$

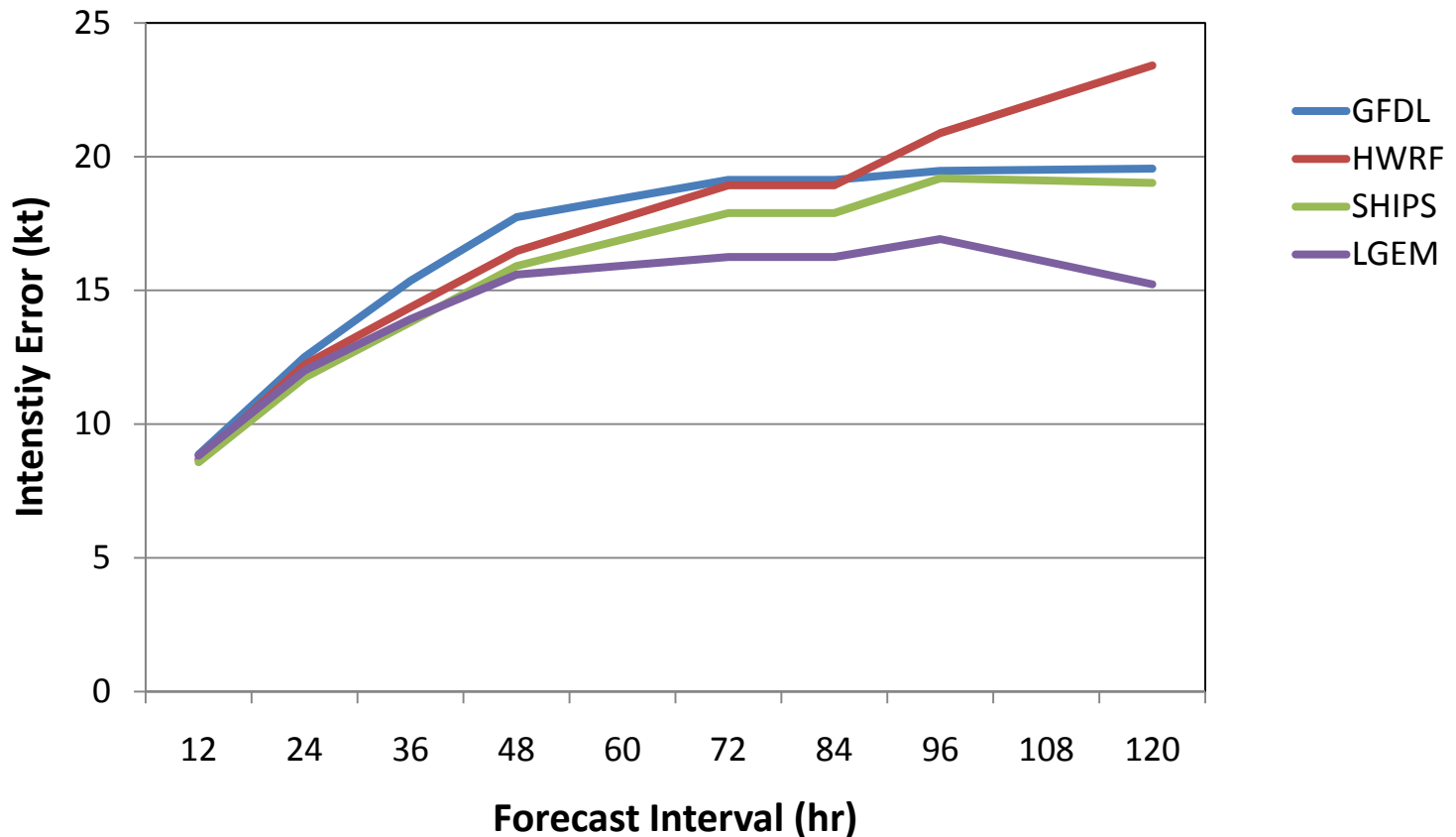
$$n = 2.6$$

$$\beta^{-1} = 39 \text{ hr}$$

$$\kappa = f(S,C)$$

V_s is MPI modified by shear and convective instability

Operational Atlantic Intensity Model Errors (2007-2009)



GFDL, HWRF are 3-D coupled ocean/atmosphere models, ~1 hr comp. time on massively parallel computer
SHIPS, LGEM are statistical-dynamical models, ~1 second on single processor Linux system

Rapid Intensity Index

- SHIPS, LGEM do not work well for “outlier” forecasts, especially rapid intensification
- RII: Operational “classification” technique developed with subset of SHIPS input
 - Discriminant analyses technique
 - Outperforms all models for rapid intensity forecasting

Previous RAMMB-CREST Collaborations on Hurricane Research

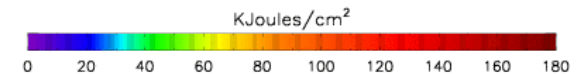
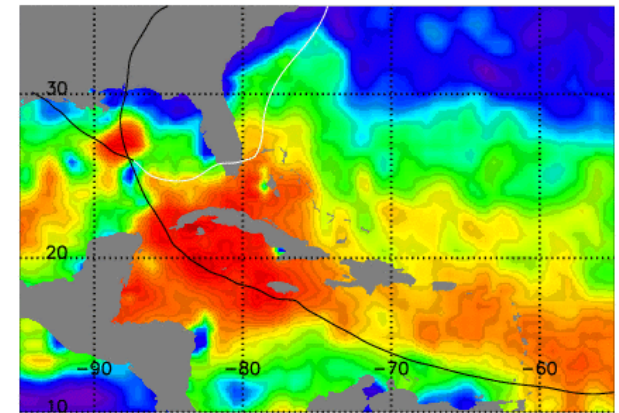
- 2005-06: Long-term hurricane environment climatology provided to Rouzbeh Nazari
- 2006-2007: Preliminary Rapid Intensification datasets sent to Michael Grossberg and Irina Gladkova for advanced statistical analysis
- 2008: GOES IR tropical cyclone dataset provided to Johnny Luo for advanced intensity estimation algorithm

Possible New RAMMB-CREST Research

Atmospheric response to ocean heat content

- Statistical relationship between OHC and intensity change in highly variable
 - Katrina case not typical
- What is the atmospheric response?
- Examine GOES evolution as storms cross warm/cold eddies
- Data needed
 - NHC best track, GOES IR and OHC archives

PRE-IVAN – Upper Oceanic Heat Content



Possible New RAMMB-CREST Research

Maximum Potential Intensity Modified By Shear

- Vertical shear is primary predictor in SHIPS, LGEM
- LGEM steady state solution is modified MPI
$$V_s = \text{Steady State} = V_{\text{mpi}}(\kappa/\beta)^{1/n} \text{ where } \kappa = \kappa(S)$$
- How does shear limit maximum intensity?
 - Test shear/MPI relationship from LGEM
 - Develop new modified MPI
 - Are there other measures of shear?
- Data needed: SST analyses, NCEP reanalysis fields, NHC best track

Possible New RAMMB-CREST Research

Advanced Statistical Techniques for Rapid Intensity Change Prediction

- Operational rapid intensity change index uses simple linear discriminant analysis, but still beats numerical model forecasts
- Can more advanced statistical techniques provide forecast improvements?
 - Nonlinear rule-based methods
 - Genetic algorithms
 - Additional predictive information through “data mining”
- Data needed: SHIPS model database (including satellite inputs), NHC best track

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Also see <http://rammb.cira.colostate.edu>