

# Estimation of Rain height from Rain Rate using Regression-based Statistical Model: Application to SeaWinds on ADEOS-II

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Committee

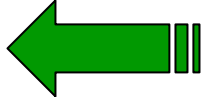
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Dr. David Braaten

# Outline



- **Introduction** 
  - Thesis Objective
- **Methodologies for Rain Height (RH) Estimation**
  - Estimation using Rain Rate (RR)
- **Regression Analysis**
  - Log-Linear Combined Regression scheme
- **Regression-based Statistical Model**
- **Validation of Statistical Model**
- **Conclusion**



# Introduction



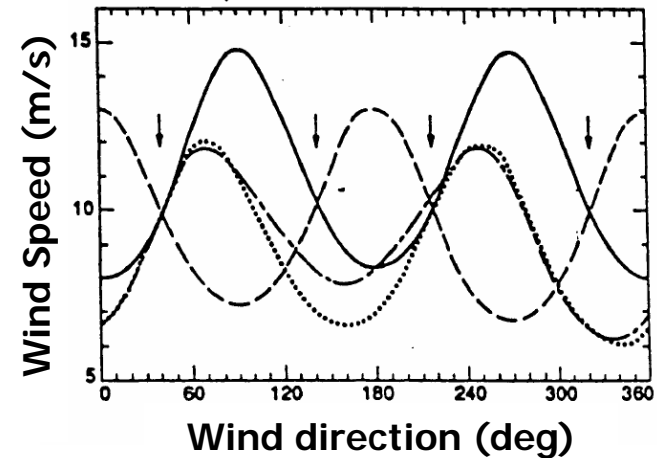
- Winds are the single largest source of momentum for the ocean surface.
- They regulate crucial coupling between air and sea, which establishes and maintains global and regional climate.
- **Wind vectors** are crucial inputs for:
  - Global numerical **weather prediction** systems
  - Monitoring unusual climatological phenomena (hurricane, El-Niño)
  - Other meteorological and oceanographic applications
- Instruments on buoys and ships can provide measurement of surface wind vectors, but they are inaccurate and their coverage is insufficient.
- Only **satellite-borne** instruments can acquire wind data with **global coverage, high spatial resolution and frequent sampling**.



# Scatterometry

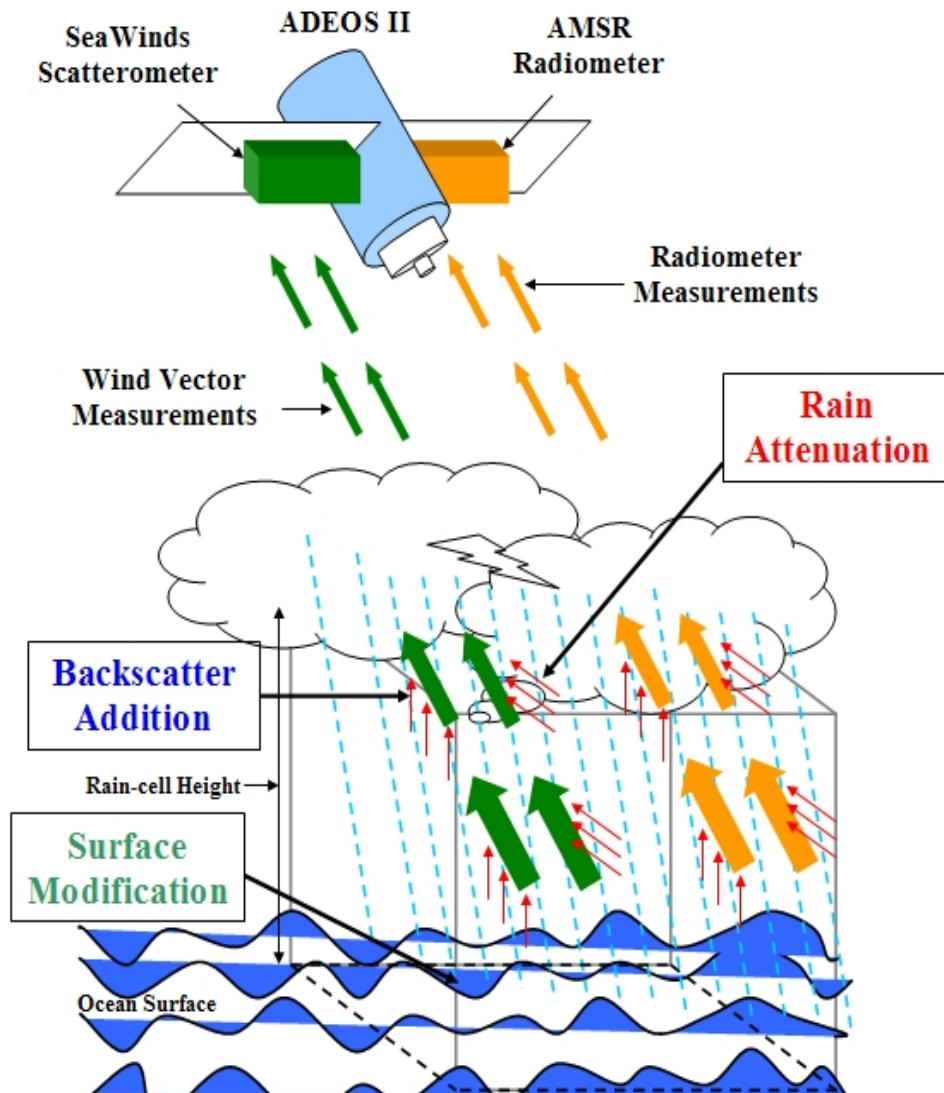


- Scatterometry for Wind Vector Measurements
  - **Scatterometer** is a microwave radar sensor, which can be used to measure ocean near-surface wind vectors for all weather conditions.
  - Radar response (surface scattering coefficient) can be related to wind vectors using **Geophysical Model Functions**.
  - Need for **Multiple** Measurements.
- SeaWinds on ADEOS-II
  - Earlier scatterometer missions: SEASAT, ERS-1/2, NSCAT, QuikSCAT.
  - NASA SeaWinds scatterometer designed at **Ku-band** (13.4 GHz)
  - SeaWinds uses a rotating dish antenna, a pencil-beam type.



# Rain Effects on Wind Data

- Rain types: **convective** and **stratiform**
- Rain effects on SeaWinds data
  - **Rain Attenuation**
  - **Backscatter Addition**
  - **Surface Modification**
- The energy backscattered by the rain can be a significant portion of the total backscatter power measured by radar.
- Most of the efforts of the research community were directed toward **flagging rain-contaminated cells** and only a few attempts were made to correct wind vector estimates for rain effects.



# Rain Effects Correction Algorithm



- Our Research
  - Develop an algorithm to correct for the rain effects, when possible
  - Develop criterion (**threshold**) for rain flag when correction impossible
- Total Received power by the scatterometer is given as the sum of attenuated surface signal and the precipitation echo.

$$P_r = P_s \tau^2 + P_v \xi$$

where,  $P_r$  is the received signal

$P_s$  is the surface signal

$\tau$  is the Total one-way transmissivity of atmosphere (<1)

$P_v$  is the power from rain volume

$\xi$  accounts for attenuation of rain echo

$$\sigma_r^0 A_s = \sigma_s^0 A_s e^{-2kR} + \eta A_s H \xi$$

Where  $\sigma_r^0$  is scattering coefficient with no correction

$A_s$  is footprint area

$\sigma_s^0$  is surface scattering coefficient

$k$  is rain attenuation constant

$H$  is Rain height - ( $A_s H$ ) is scattering volume

$R$  is slant range through the rain also related to rain height

$\eta$  is volume scattering coefficient



# Rain Effects Correction Algorithm (contd.)



## ■ Correction Algorithm

- To correct the signal, we invert the previous equation

$$\sigma_s^0 = (\sigma_r^0 - \eta H \xi) e^{2kR}$$

- The correction of  $\sigma_s^0$  from rain effects requires the following inputs:
  - Reflectivity factor ( $Z$ ) related to volume scattering coefficient
  - Attenuation constant ( $k$ )
  - Rain height ( $H$ )

- The reflectivity factor and attenuation constant can be obtained using empirical relationships with rain rate.

$$k = \alpha(RR)^\beta \quad Z = a(RR)^b$$

- $\alpha$  and  $\beta$ , and  $a$  and  $b$  (different for stratiform and convective), are found in the literature.
- The availability of rain rate from **AMSR radiometer** in ADEOS II made this approach possible for ADEOS-II, and presumably will do so for future missions.



# Thesis Objective


- **Rain height** (RH) or storm height as used in our analysis, is the height of the top of the rain column above the mean sea level – first echo measured by radar
- Rain height is critical to get path length for attenuation and volume for backscatter.
- Unfortunately, Rain height is **not available** in any of the instruments carried by ADEOS-II.
- **Thesis Objective**
  - To identify a **methodology to estimate rain cell height** over oceans using statistics or from any available ADEOS-II instrument's output
  - This estimated rain cell height should be valid over different seasons, rain types and regions of the globe.





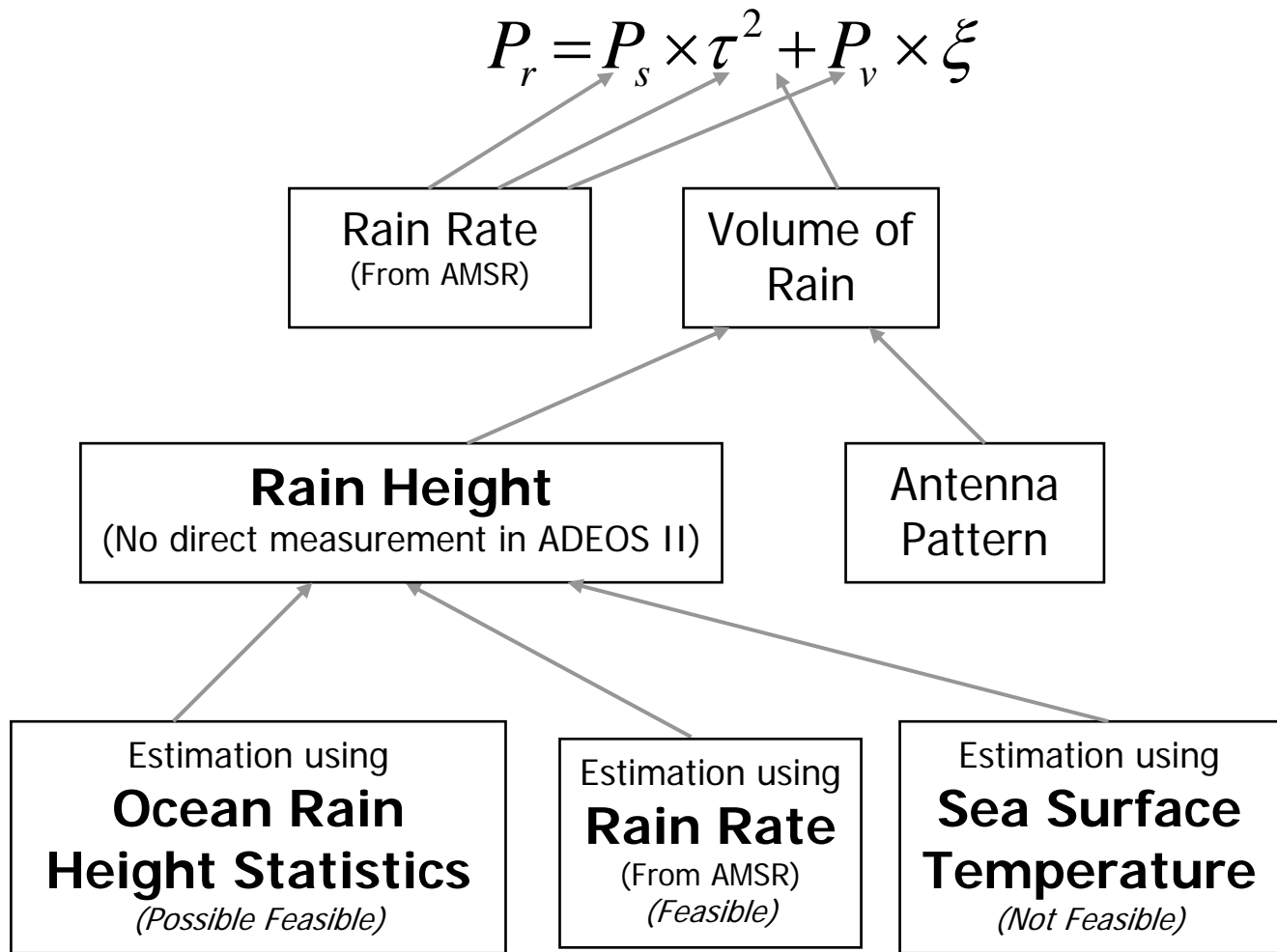
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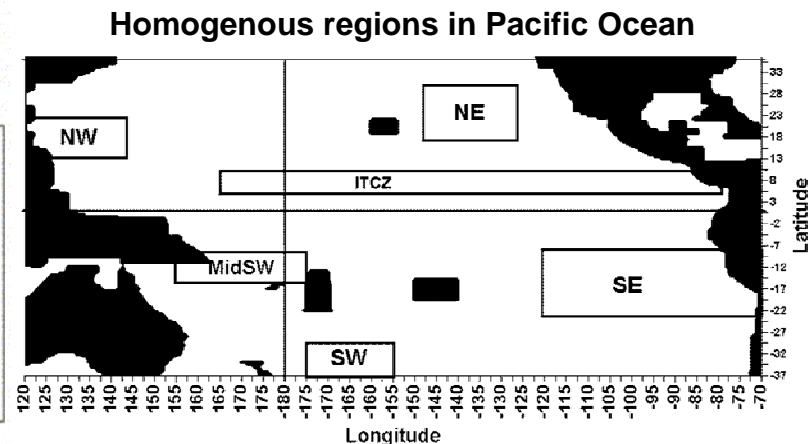
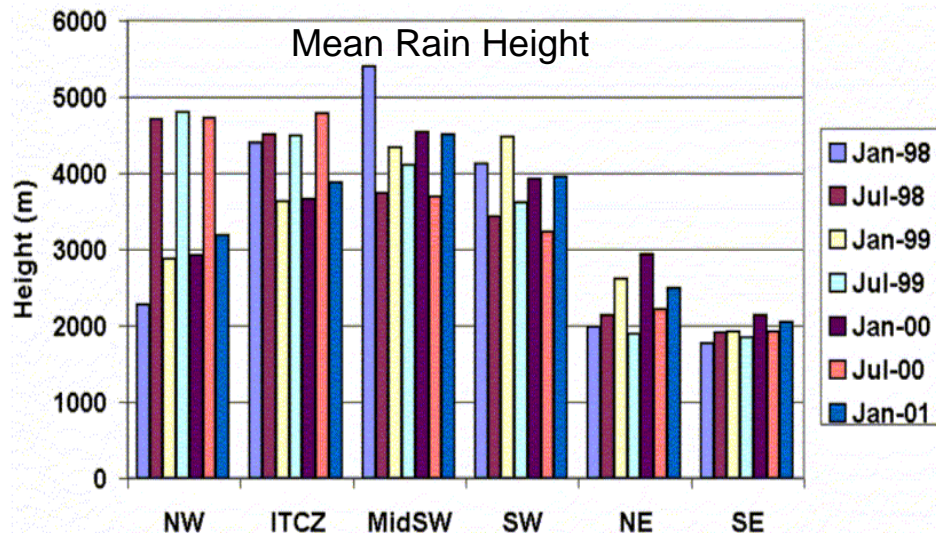
# Methodologies for RH estimation



# Climatological Rain Height

## Rain Height Statistics

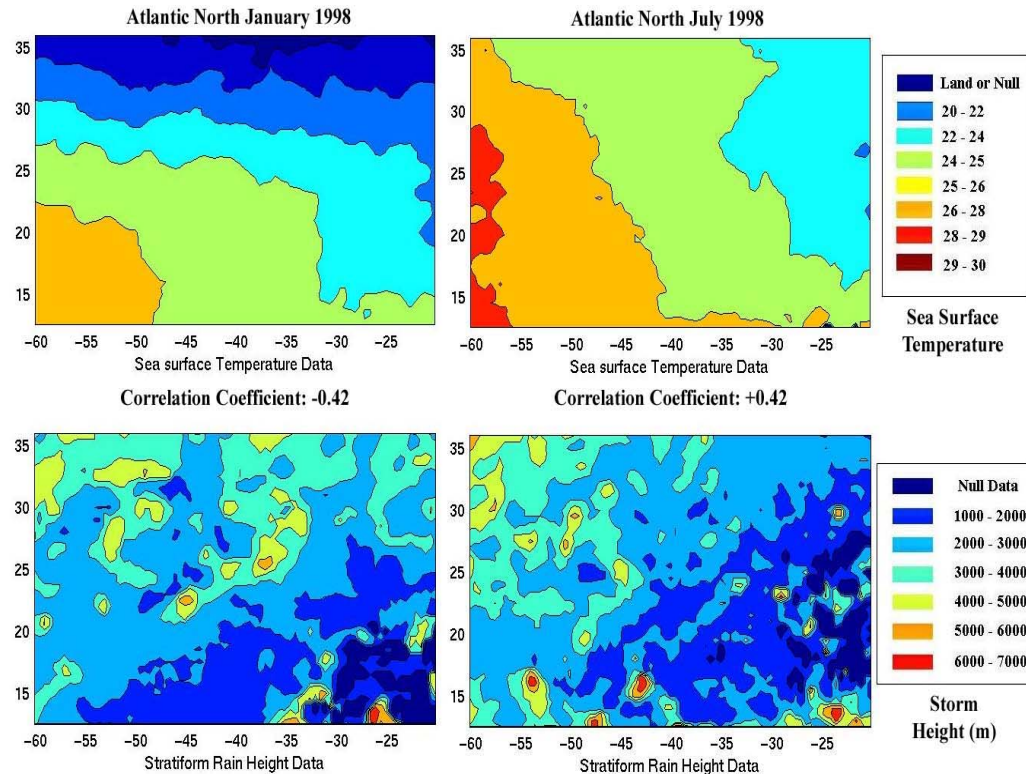
- Conventional statistics from weather services do not provide this information.
  - We attempted to analyze Level-3 TRMM precipitation radar monthly averaged rain-height data and generate mean rain-height tables.
  - Oceanic regions are subdivided into homogenous rain height regions to account for variation in regional ocean climate.
- This approach provides only a **crude** statistical measure of rain height.
- Another drawback of this approach, is the **limited latitude coverage** of TRMM ( $\pm 35^\circ$  Latitude). Hence, mean rain-height tables cannot be developed for higher latitudes.



# From Sea Surface Temperature



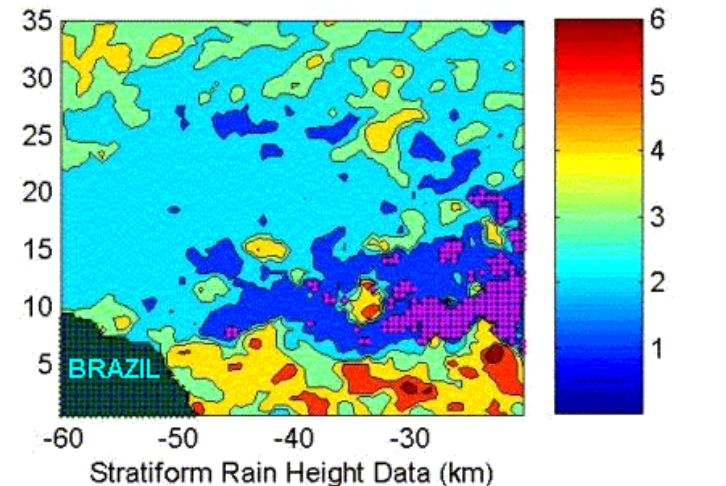
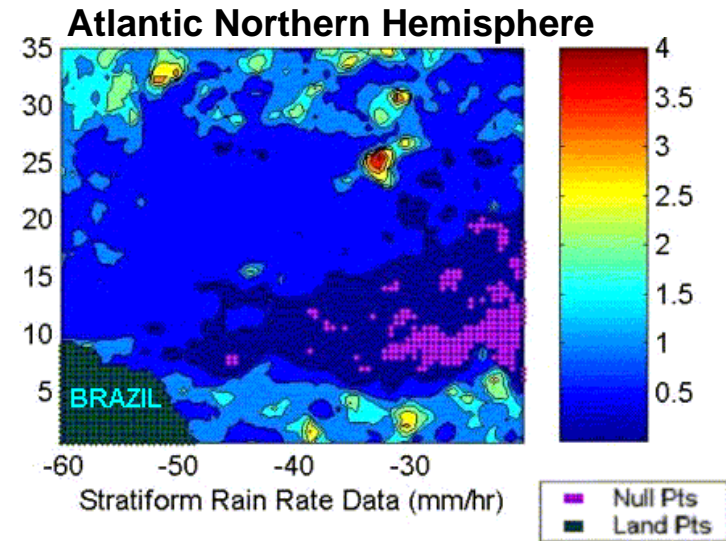
- Estimation of rain height from Sea-surface Temperature
  - TRMM PR rain height products and AVHRR sea-surface temperature (SST) data are correlated.
  - Although some regions show high positive correlations with SST, others sometimes correlate negatively, so the SST based rain-height estimation approach **cannot be used throughout the world's ocean**



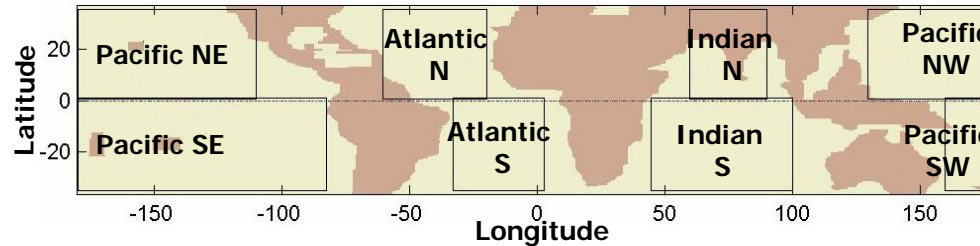
# Estimation using Rain Rate



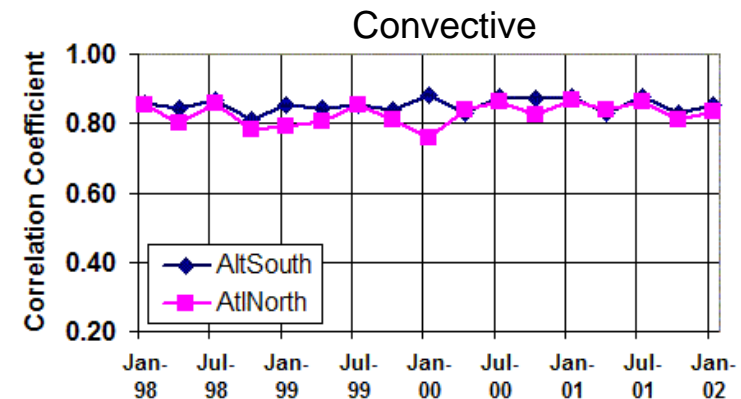
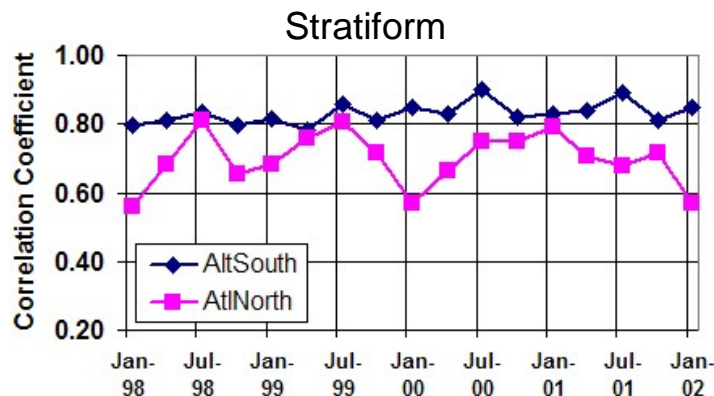
- Correlation between Rain-rate and Rain-height data
  - TRMM precipitation radar level-3 monthly averaged rain-rate and rain-height products are used in this study.
- Basic **Data processing** includes:
  - Same rainfall type is chosen from both the datasets (RR & RH).
  - Spurious point removal based on rain-count data.
  - **Land Mask** – All islands in major groups combined to single areas and the widely dispersed small islands are ignored.
  - Null-data and land masked points are omitted.



# Estimation using Rain Rate



- Correlation is done on selected regions on quarterly basis
- Correlation coefficients are tabulated for further analysis
- Correlation coefficients are very high for most of the regions, seasons and rain types.
- The **strong correlation** between rate and height over the ocean demonstrates the feasibility of estimating rain height using rain rate.



# Outline

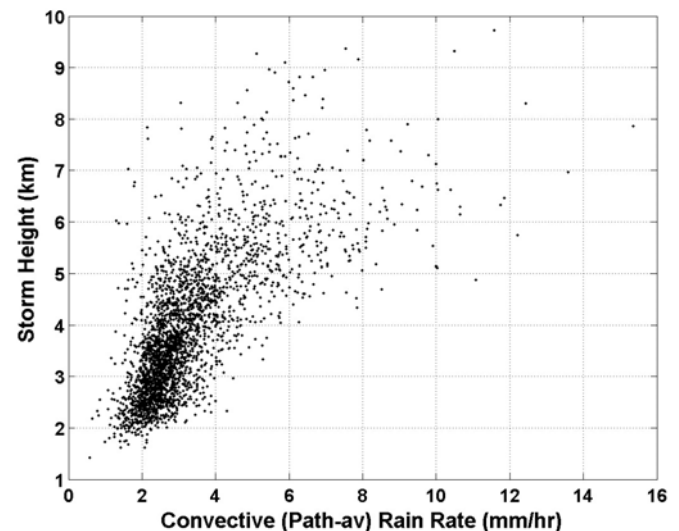
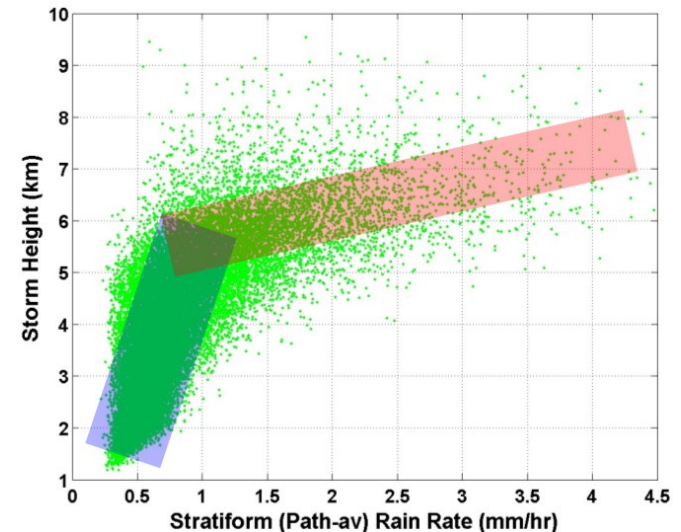


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# RR vs. RH scatter

- Mathematical dependency of rain height to rain rate is analyzed using regression analysis.
- TRMM PR's rain height and rain rate products are used in this analysis.
- Data processing of TRMM rain products are similar to the previous section.
- RR vs. RH scatter exhibits **two different trends** for different rain-rate regimes.
  - Lighter rain: RH vs. RR relation is steep
  - Heavier rain: scatter almost saturates
- RR vs. RH scatter exhibits similar trend for all rain types, and regional and hemispherical based data.





# Objective of Regression methods

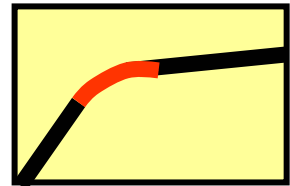


- **Good fit over all RR**

- Importance in Lower RR section because of large number of scatter points.

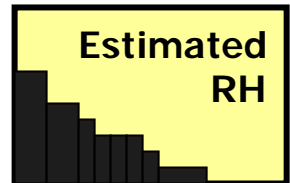


- **Smooth Transition** from one trend to another to maintain consistency.



- **Estimated RH distribution**

- The estimated rain-height distribution should be consistent with the original rain-height distribution.



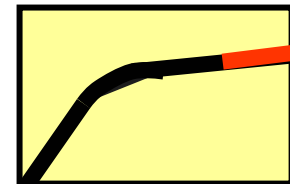
- **Universal Regression scheme**

- It is important to have a regression scheme that applies to various regions, seasons, and rain types because, for simplicity, one should have a single regression model for final prediction of storm height from rain rate.

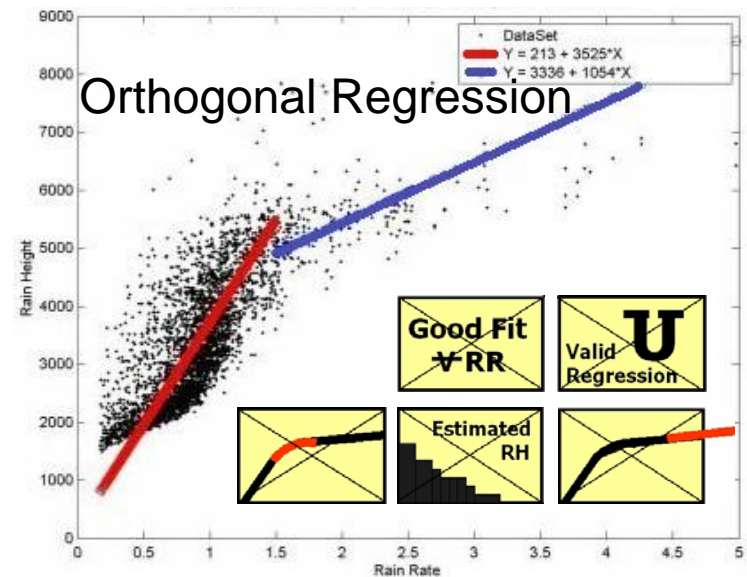
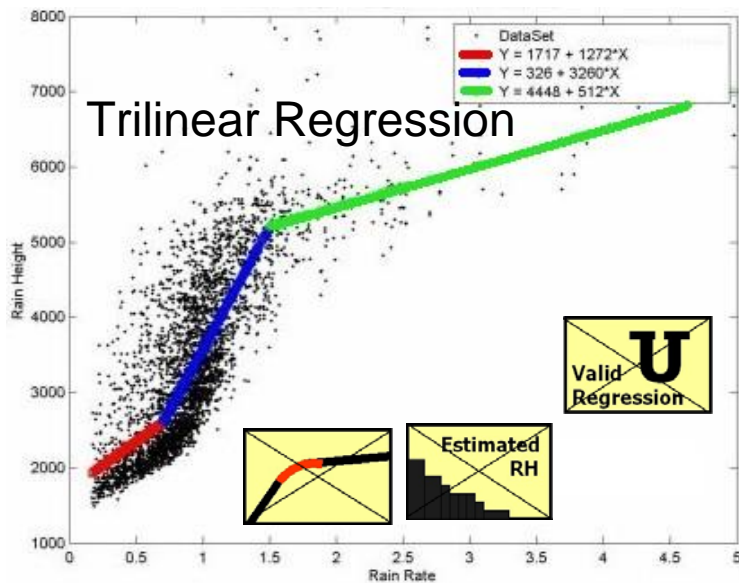
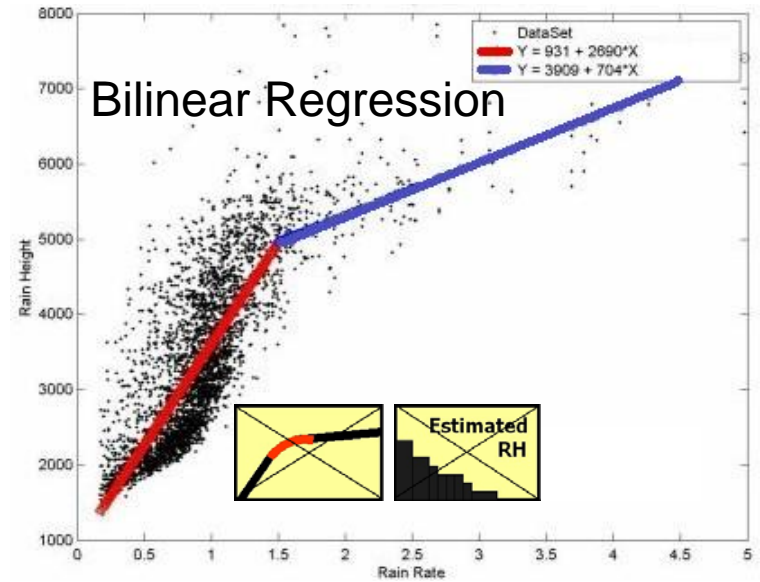
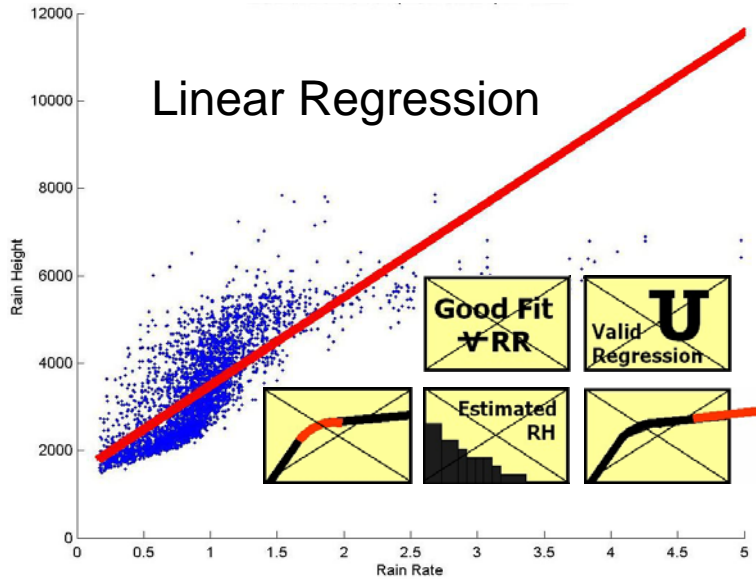


- **Linear extrapolation**

- Mathematical relationship should be extendable.

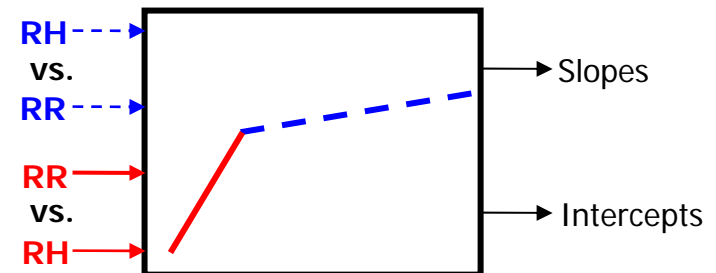
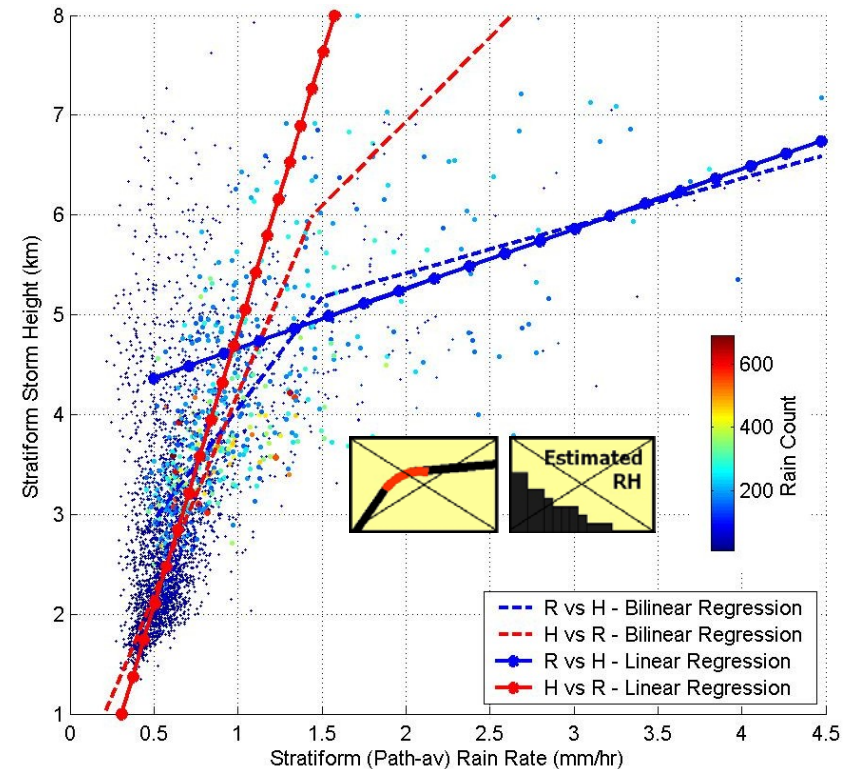


# Linear Regression Schemes



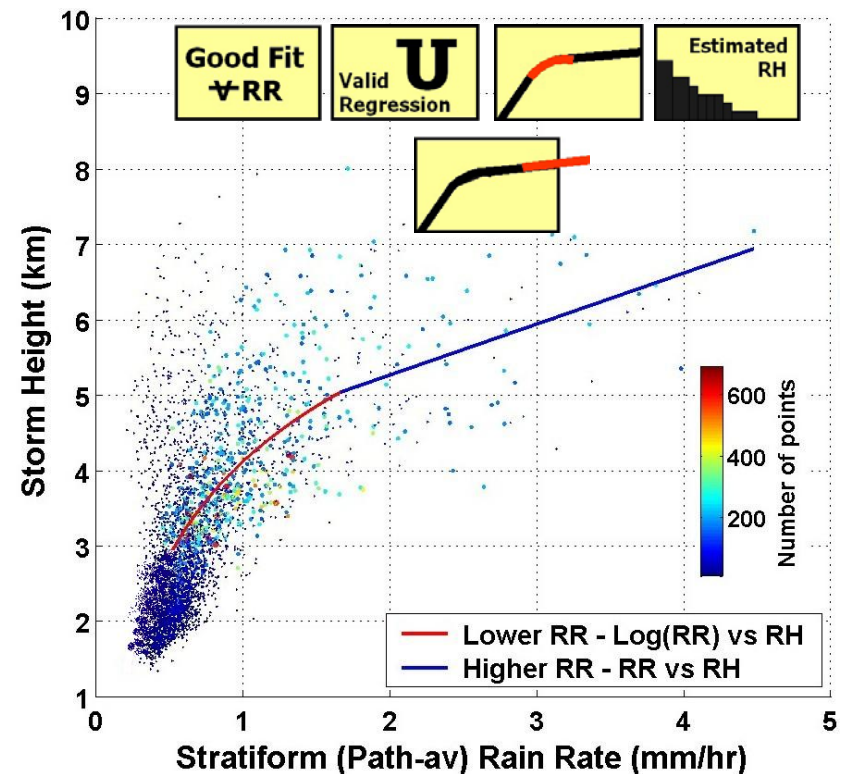
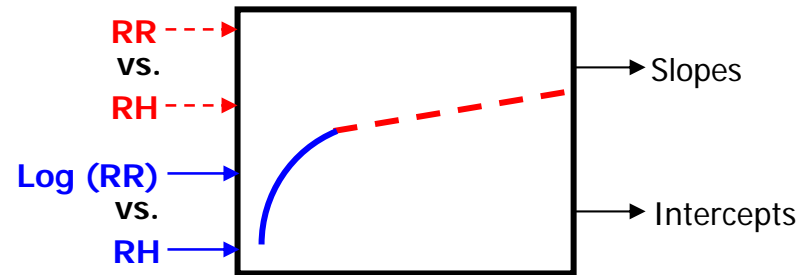
# Combined Linear

- Bilinear regression lines tend to **influence each other**, because of the additional constraints applied to make the lines meet at a point.
- Two linear regression lines:
  - RR vs. RH regression for lower rain rates (red solid)
  - RH vs. RR regression for higher rain rates (blue solid)
  - Individual linear regression lines seem to fit the scatter better
- **Combined Linear** regression
  - Two linear regression lines extended to all RR forms  
Combined Linear regression
- Drawback: **sharp transitions** between lower and higher regression lines



# Log-Linear Combined

- *Log-linear Combined* regression is a modification to the *Combined Linear* regression to overcome the drawbacks.
- In *Log-Linear Combined* regression, two linear regression lines:
  - Log(RR) vs. RH regression for lower rain rates.
  - RR vs. RH regression for higher rain rates.
- *Log-Linear combined* provides smooth transition between different trends and good fit for all regions and rain types.
- *Log-Linear Combined* is selected as the best regression scheme to relate rain rate and rain height.



# Statistical Significance

- Statistical measures
  - **Standard Error of the estimate** – square root of the average squared error in estimation
  - **Goodness of fit** is a measure of the extent to which the total variation of the dependent variable is explained by the regression scheme
- *Log-Linear Combined* regression shows best result compared to all other regression schemes
- To visualize the accuracy of our regression fit over all rain-rate values, we segmented the RR vs. RH scatter and estimated standard error in height for each segment. *Log-Linear Combined* regression showed consistently low standard error in height for all rain-rates segments

| Regression schemes                      | Standard Error of Estimate (km) | Goodness of fit $R^2$ |
|---|---------------------------------|-----------------------|
| Linear Regression                       | 2.5                             | 0.4                   |
| Bilinear Regression                     | 1.3                             | 0.52                  |
| Weighted Bilinear Regression            | 1.2                             | 0.55                  |
| Log(RR) vs. RH Regression               | 1.4                             | 0.60                  |
| 'Combined Linear' Regression            | 0.93                            | 0.89                  |
| <b>'Log-Linear Combined' Regression</b> | <b>0.87</b>                     | <b>0.93</b>           |

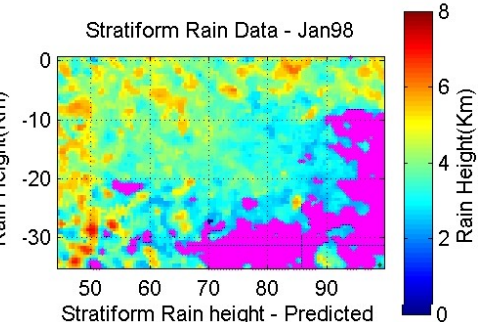
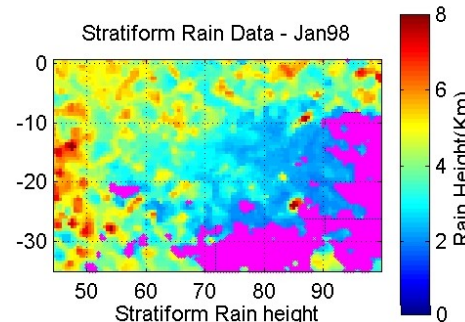
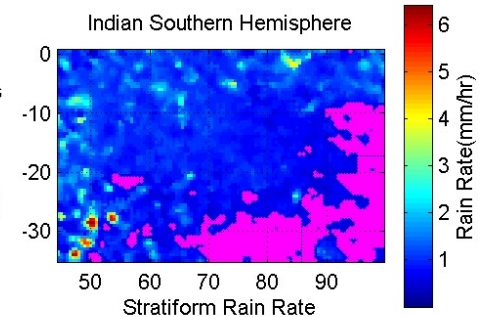


# Hypothesis T-test

- Hypothesis Paired-t test was used to find the closeness of the original TRMM rain height to that of the estimated rain height obtained from the regression model.
  - The null hypothesis in this case is that the means of the two rain height data are equal.
  - After analyzing the data based on T-statistics, the t-test produces a result to either “reject the null hypothesis” or “do not reject the null hypothesis”
  - The p-value result from the t-test is the probability of the observing the given sample result under the assumption that the null hypothesis is true.
- 
- The p-value for this test over Indian Southern Hemisphere is given as **0.9284**.
  - Higher p-value implies **higher probability** for the null hypothesis to be true.
  - Similar results are observed for rain data from other stratiform regions.
  - Convective rain data typically results in lower p-value, compared to the stratiform case.

Two-Sided T-test : Hypothesis test  
Compares the averages of the two samples  
Null Hypothesis is : "means are equal"

H=0 Significance level:  $\alpha = 0.05$   
(if H=0=>Do not reject Null Hypothesis @ $\alpha$ )  
(if H=1=>Reject Null Hypothesis @ $\alpha$ )  
Pvalue = 0.928384



# Regional vs. Hemispherical based RR estimates



- Predicted rain-height from the **regional-based** RR vs. RH scatter is compared to **hemispherical-based** RR vs. RH scatter (t-test).
- The table shows **higher confidence in northern and southern hemisphere** scatter based rain-height prediction compared to region data based rain-height prediction.
- This result is very significant as one can **omit the regional data** based analysis for the rest of the statistical model development.
- Similar results were observed for convective case.

Stratiform

| Region                     | p-value     |
|----------------------------|-------------|
| Indian north               | 0.80        |
| Atlantic north             | 0.91        |
| Pacific northeast          | 0.78        |
| Pacific northwest          | 0.75        |
| <b>Northern Hemisphere</b> | <b>0.90</b> |
| Indian south               | 0.93        |
| Atlantic south             | 0.92        |
| Pacific southeast          | 0.89        |
| Pacific southwest          | 0.85        |
| <b>Southern Hemisphere</b> | <b>0.94</b> |

Convective

| Region                     | p-value     |
|----------------------------|-------------|
| Indian north               | 0.73        |
| Atlantic north             | 0.75        |
| Pacific northeast          | 0.77        |
| Pacific northwest          | 0.69        |
| <b>Northern Hemisphere</b> | <b>0.75</b> |
| Indian south               | 0.74        |
| Atlantic south             | 0.72        |
| Pacific southeast          | 0.69        |
| Pacific southwest          | 0.70        |
| <b>Southern Hemisphere</b> | <b>0.79</b> |



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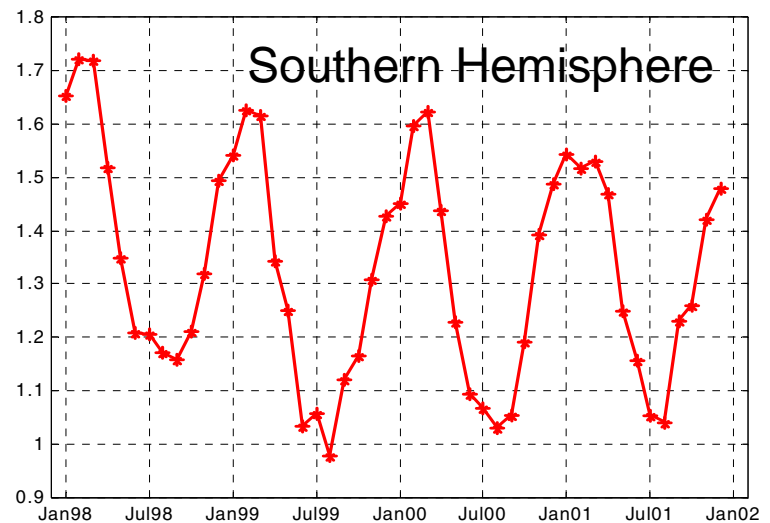
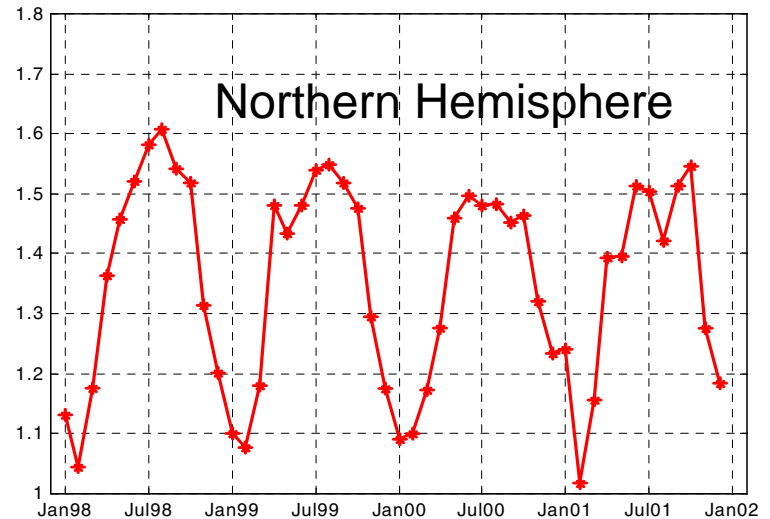
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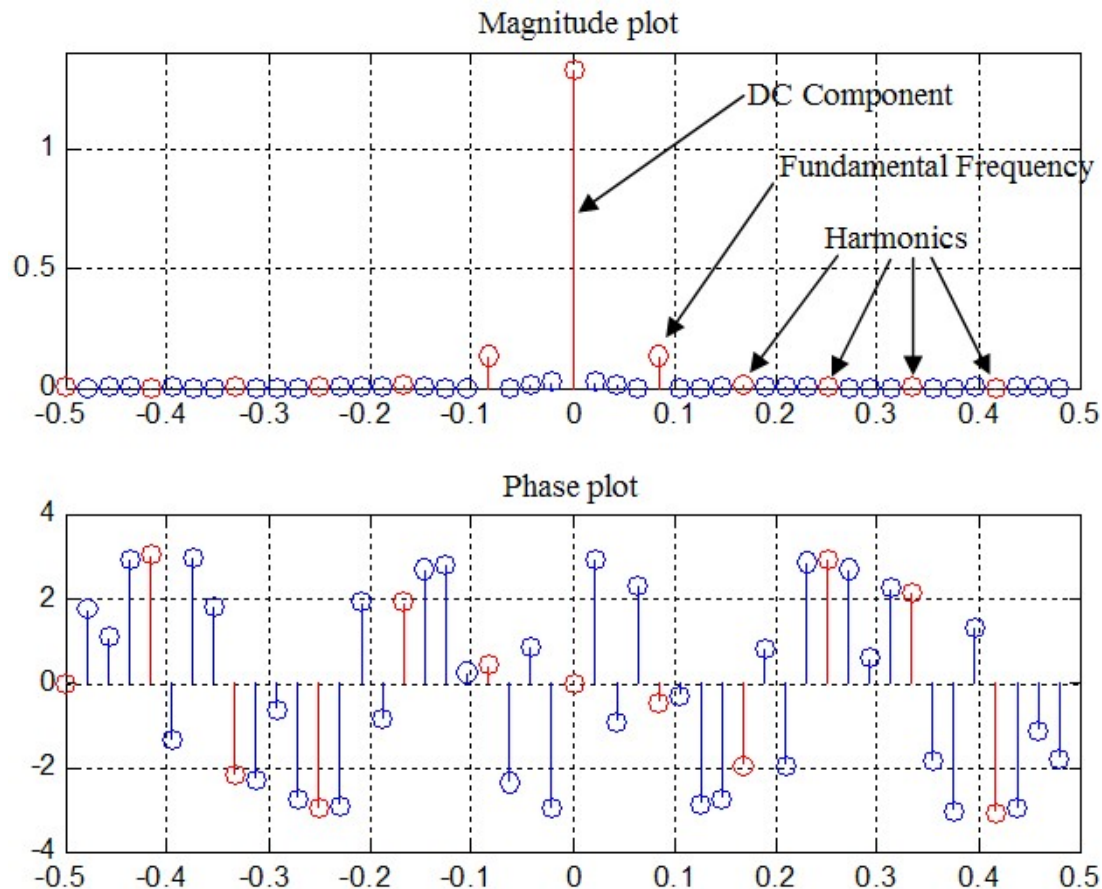
# Slope and Intercept Statistics

- Slopes and intercepts of *Log-Linear Combined* regression schemes are **tabulated**.
- Values of slopes from the regression fits clearly show **sinusoidal trend over seasons** with approximately one year as the period of each cycle.
- The northern and southern hemisphere slope curves are approximately out of phase.
- The seasonal trend in slopes and intercepts statistics have to be generalized to **predict them for any future month**.
- Mean-value substitution not possible since they affect the estimated rain-height.



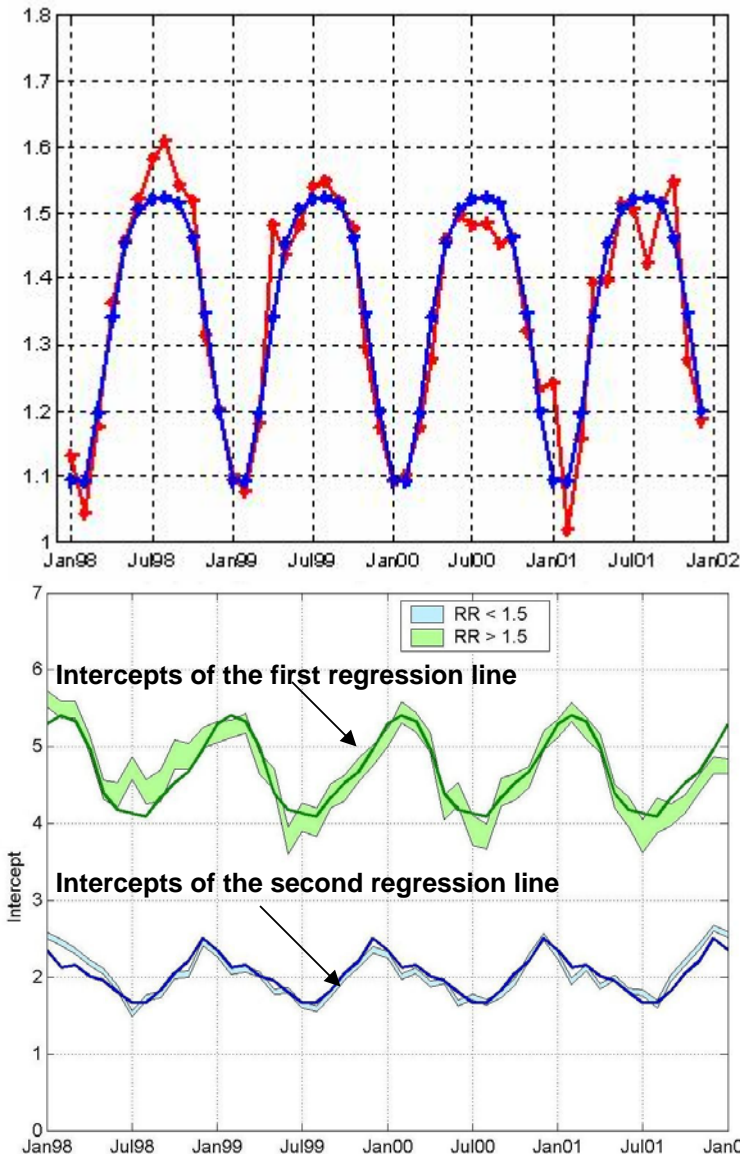
# Fourier Analysis

- **Fourier Analysis** can capture the seasonal trend exhibited by the slope and intercept statistics.
- Fourier coefficients can be obtained by analyzing the spectrum of the slope and intercept curves.



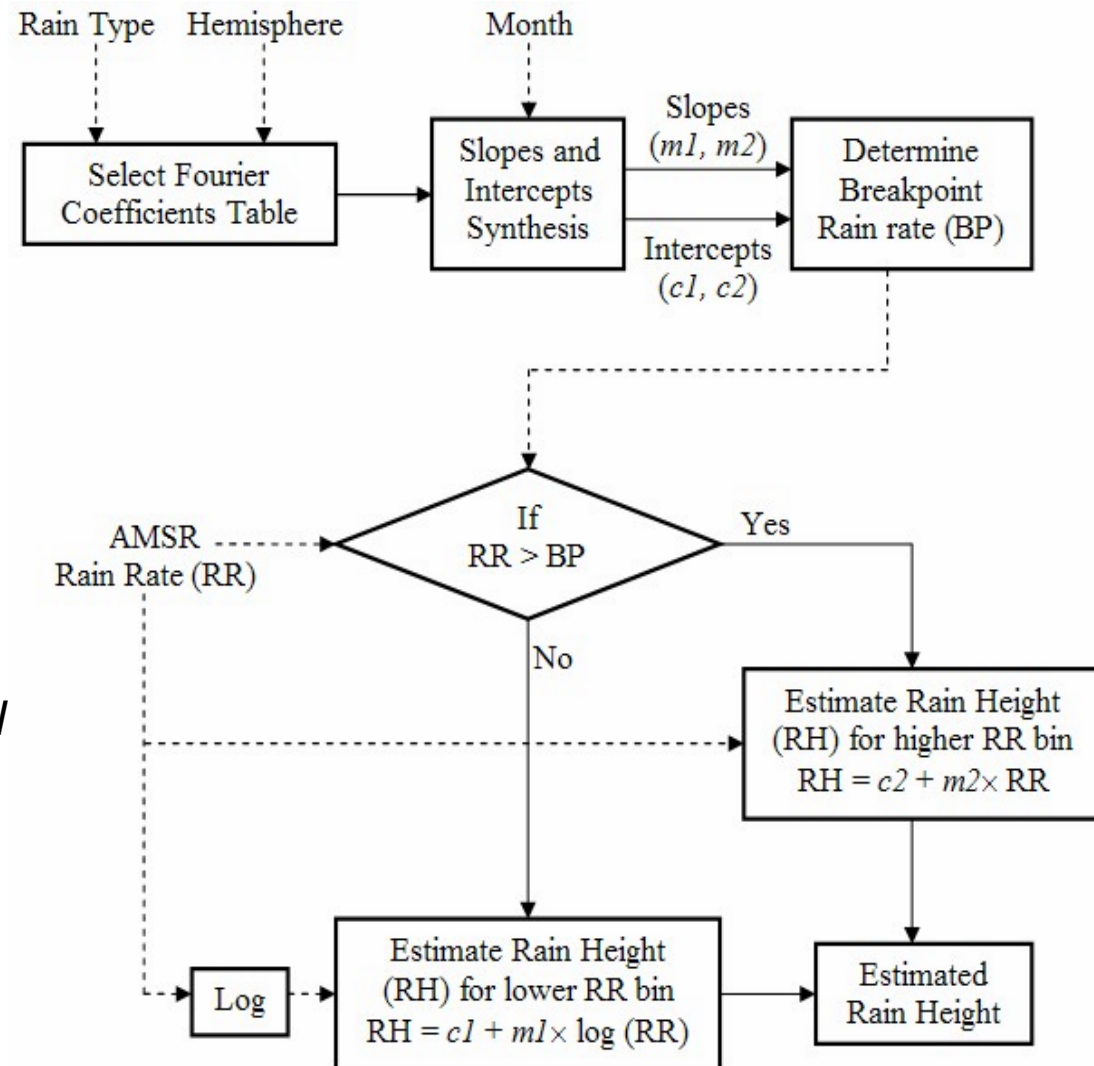
# Fourier Synthesis

- After analyzing various slopes and intercept statistics, we concluded that the use of **DC, fundamental and two harmonics** are sufficient to regenerate the basic sinusoidal trend.
- Fourier synthesized curves along with the original slope statistics are shown.
- Using Fourier coefficients, the slopes and intercepts of the *Log-Linear Combined* regression scheme can be predicted for given month in the future.
- **Table of Fourier coefficients** are the final output of this analysis. Only four Fourier coefficient tables are required in the *Statistical Model*. They are:
  - Stratiform Northern Hemisphere
  - Stratiform Southern Hemisphere
  - Convective Northern Hemisphere
  - Convective Southern Hemisphere



# Regression-based Statistical Model

- Slopes and intercepts are generated from the selected Fourier coefficient table based on rain type, hemisphere and month.
- The slopes and intercept values of the two regression lines are used to find the **breakpoint rain-rate**. This would split the scatter into two rain-rate sections.
- *Regression-based Statistical Model* uses the predicted slopes and intercepts of the *Log-Linear Combined* regression scheme to estimate rain height from AMSR rain rate.
- **Final output of this thesis**



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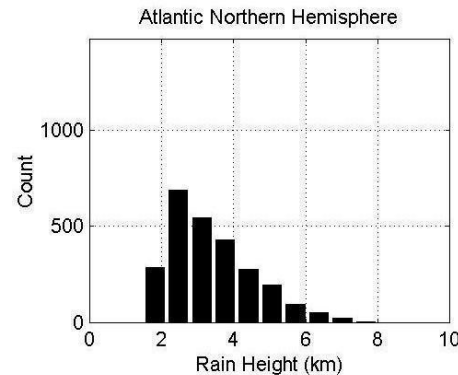


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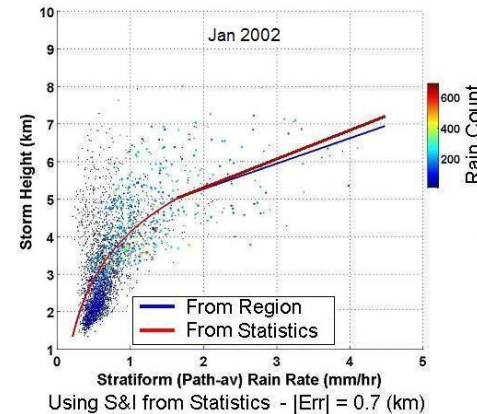
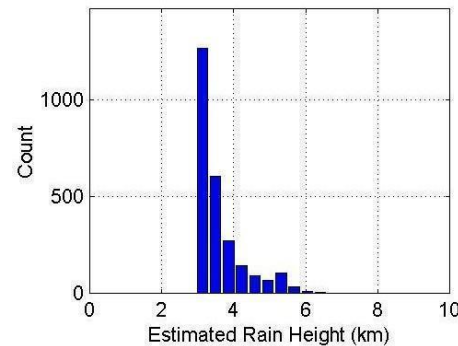


# RR estimates from statistical model

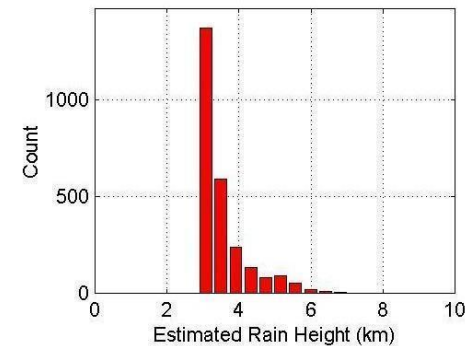
- The rain height estimates obtained from *Log-Linear Combined* regression fits are compared with the rain height estimates derived from the statistical model.
- The **Kolmogorov-Smirnov** (ks-test) goodness-of-fit hypothesis test is used to compare the estimated rain-height distributions. The test results shows “not to reject the null hypothesis” and with a p-value of **0.92**
- This comparison highlights the effectiveness of the statistical model in predicting the slopes and intercepts of the regression lines and the subsequent rain-height estimation.



Using S&I from Region - |Err| = 0.72 (Km)



Using S&I from Statistics - |Err| = 0.7 (km)



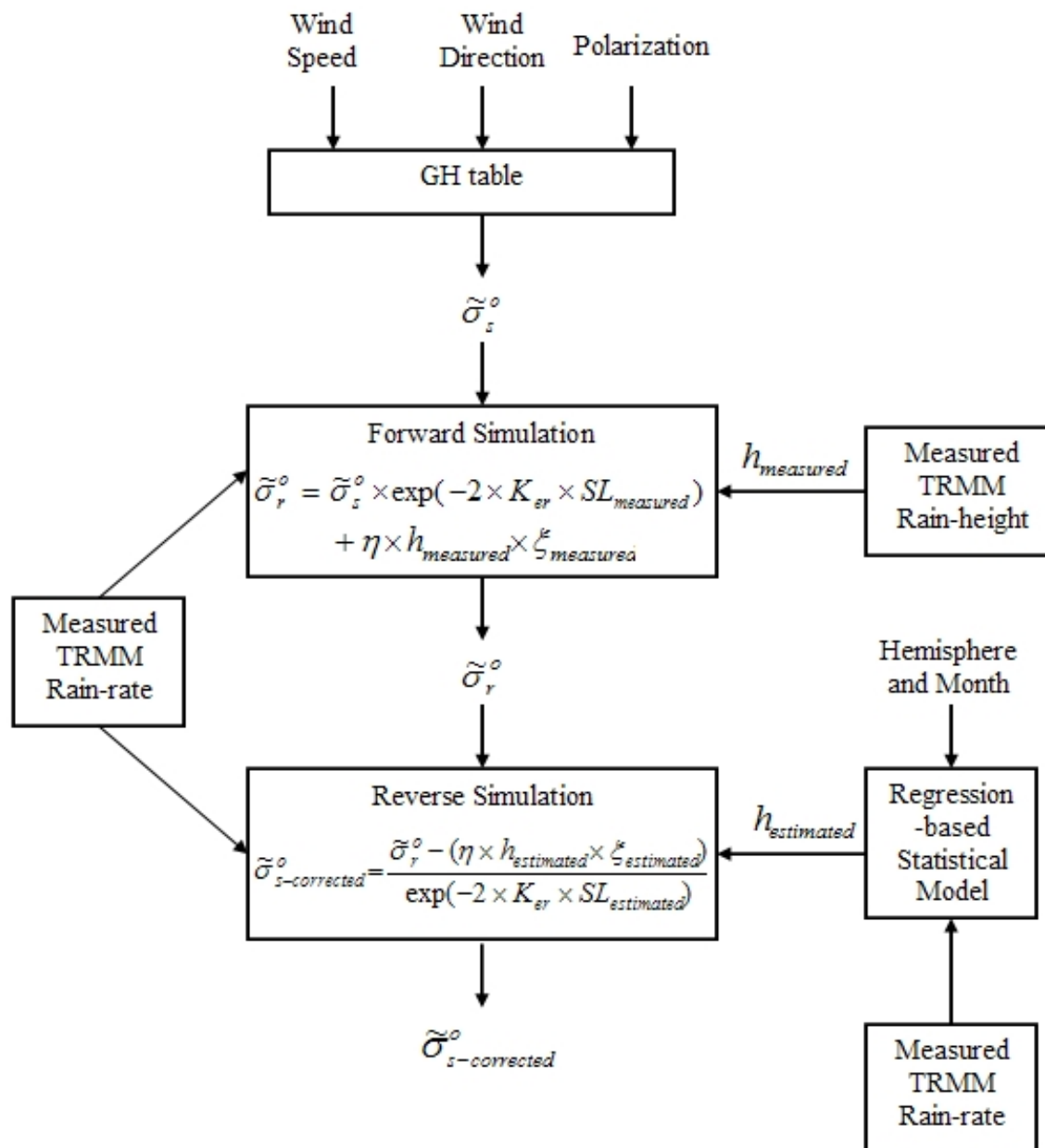
# Simulation Procedure



- Rain-height estimates from the *statistical model* will be in error compared to the actual rain-height values.
- Simulations were performed to determine the nature of the errors in  $\sigma_s^0$  with the use of the estimated rain-height from the *regression-based statistical model*.
- **Forward Simulation**
  - Estimates the  $\sigma_r^0$  as measured by the SeaWinds instrument.
  - The forward simulation models the rain effects in the SeaWinds received signal with the use of **measured TRMM rain-height** values  $h_{measured}$ .
- **Reverse Simulation**
  - The reverse simulation tries to correct rain effects in the  $\sigma_r^0$  to estimate the corrected  $\sigma_s^0$
  - The **estimated rain-height**  $h_{estimated}$  from the *regression-based statistical model* is used in the reverse simulation to model the effects of rain.



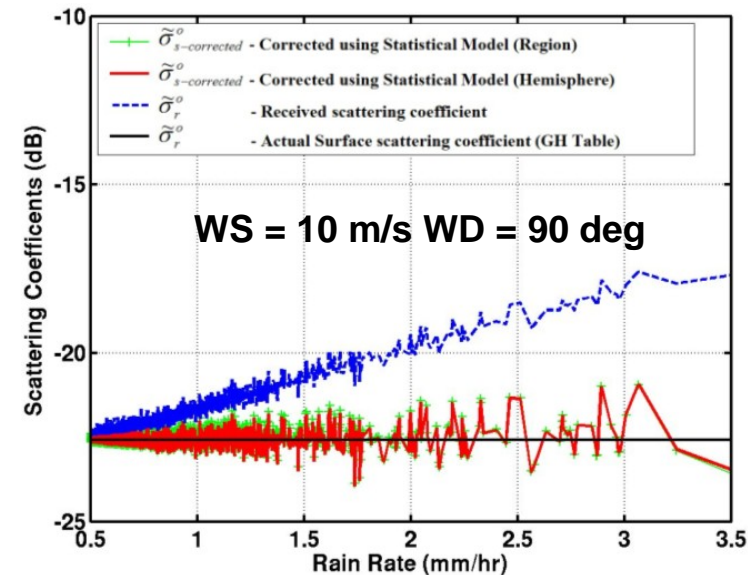
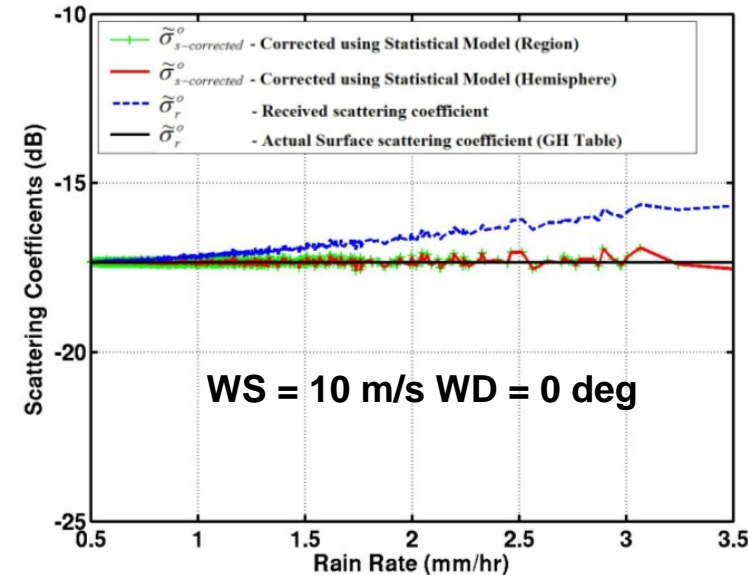
# Simulation Procedure (contd.)





# Simulation Results

- Simulations were performed for different combinations of wind speed, wind direction and polarization.
  - The errors in the corrected surface scattering coefficient are greater at **higher rain rates** – due to the large spread of data points for those rain rates.
  - Errors between the corrected and true scattering coefficients were **greater at lower wind speeds** than at higher wind speeds. Also we observe that the error increases as the look direction approaches crosswind.
- Corrected scattering signal obtained from statistical model developed based on regional (green) and hemispherical (red) data are very close to each other.
- The *regression-based statistical model* was found to be effective for rain-height estimation for most cases.
  - However, for some high rain-rates and other wind conditions, correction is not possible – **rain flags**



# Conclusion



- SeaWinds scatterometer measurements are corrupted by rain attenuation and backscatter.
- SeaWinds on ADEOS-II had AMSR available to aid correction for rain effects.
- To estimate volume backscatter from rain, must have **estimate of rain height** - Unavailable in ADEOS-II.
- Our study of TRMM RR and RH showed good correlation.
- Due to the nature of the RR vs. RH scatter, **Log-Linear Combined** regression (log(RR) vs. RH for lower RR and RR vs. RH for higher RR) was selected.
- To apply this regression-based relationship for all seasons, the slope and intercept statistics were Fourier analyzed to capture the seasonal trends.
- Finally, we proposed the **regression-based statistical model**, which estimates the rain height from rain rate for a given hemisphere, season, and rain type.
- Simulation results showed good rain-effect correction for surface scattering coefficient for lower rain-rates, validating the use of statistical model in the *rain effect correction algorithm*.
- Although **ADEOS-II failed**, the methods for correction will be useful in future satellites carrying both wind-vector scatterometers and microwave radiometers



**Thank you**