



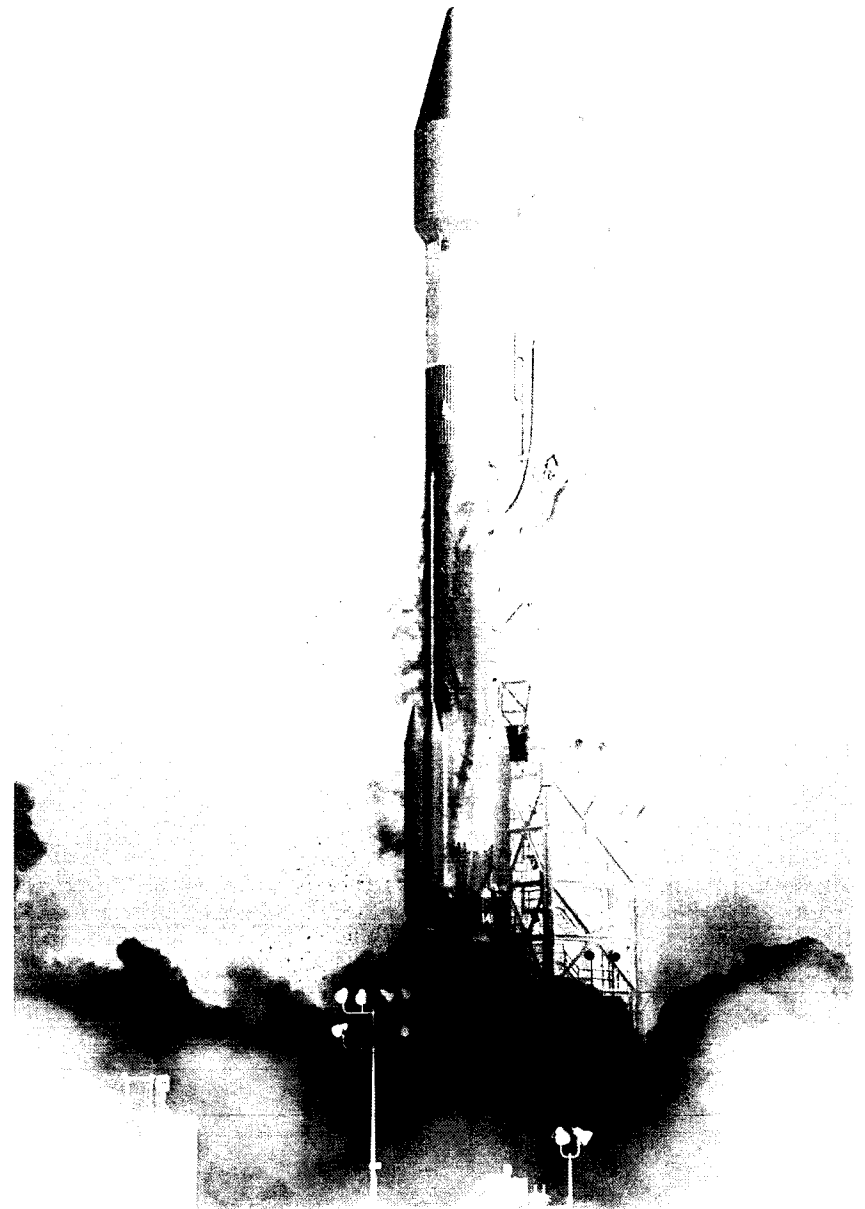
National Aeronautics and  
Space Administration

Jet Propulsion Laboratory  
California Institute of Technology  
Pasadena, California

# Data Fusion for Earth Science Remote Sensing

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*Launch of the Terra satellite, December 19, 1999*



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# Outline



*Data collected from multiple sensors on Terra.*

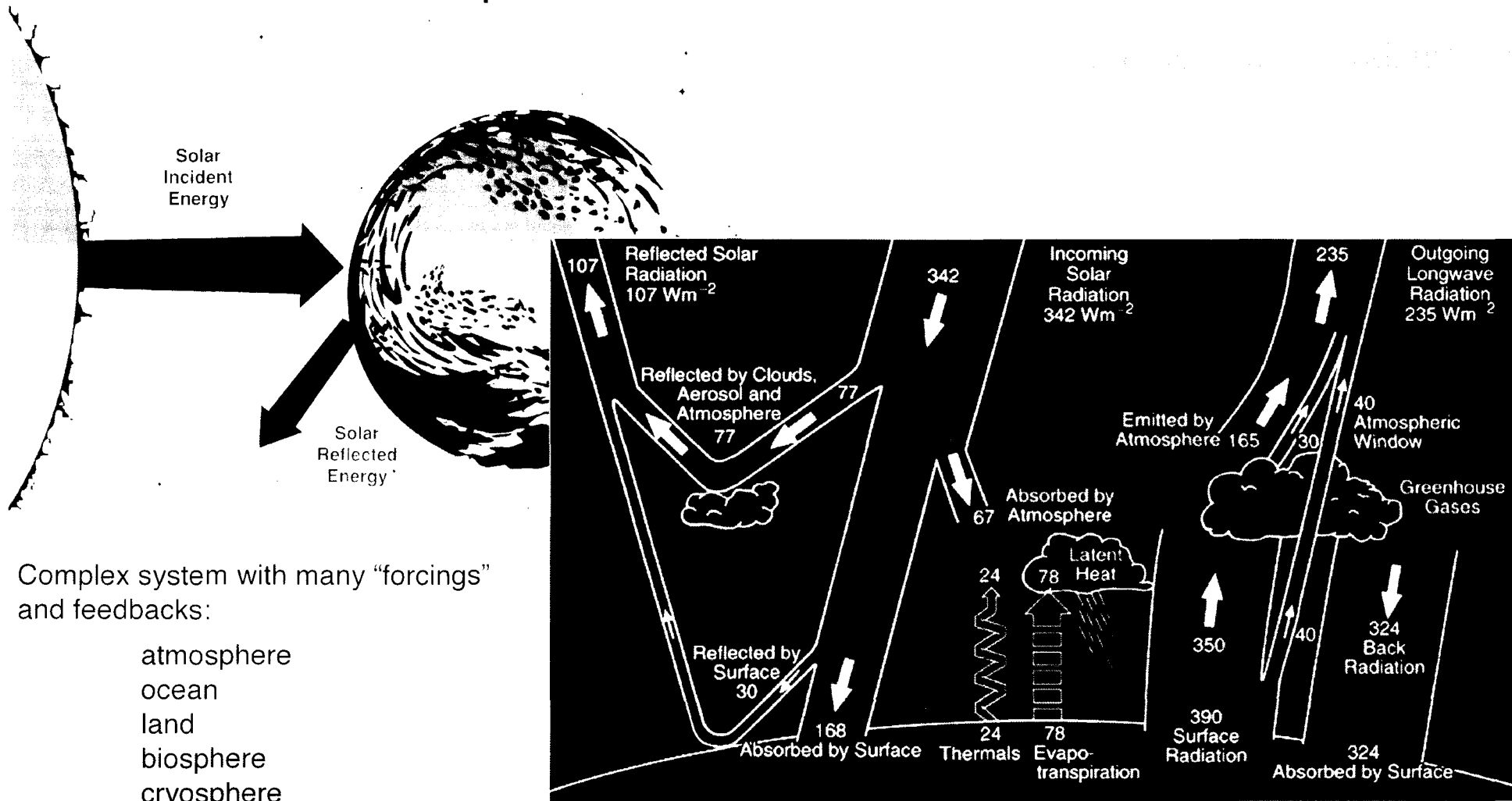
- Introduction: NASA's Earth science mission.
- The Earth Observing System.
- Satellite data acquisition and processing.
- The challenge of data fusion.
- A framework.
- Conclusion: your ideas?



# Introduction

NASA's Earth science mission: *To understand and protect our home planet*

## Earth Radiation Components

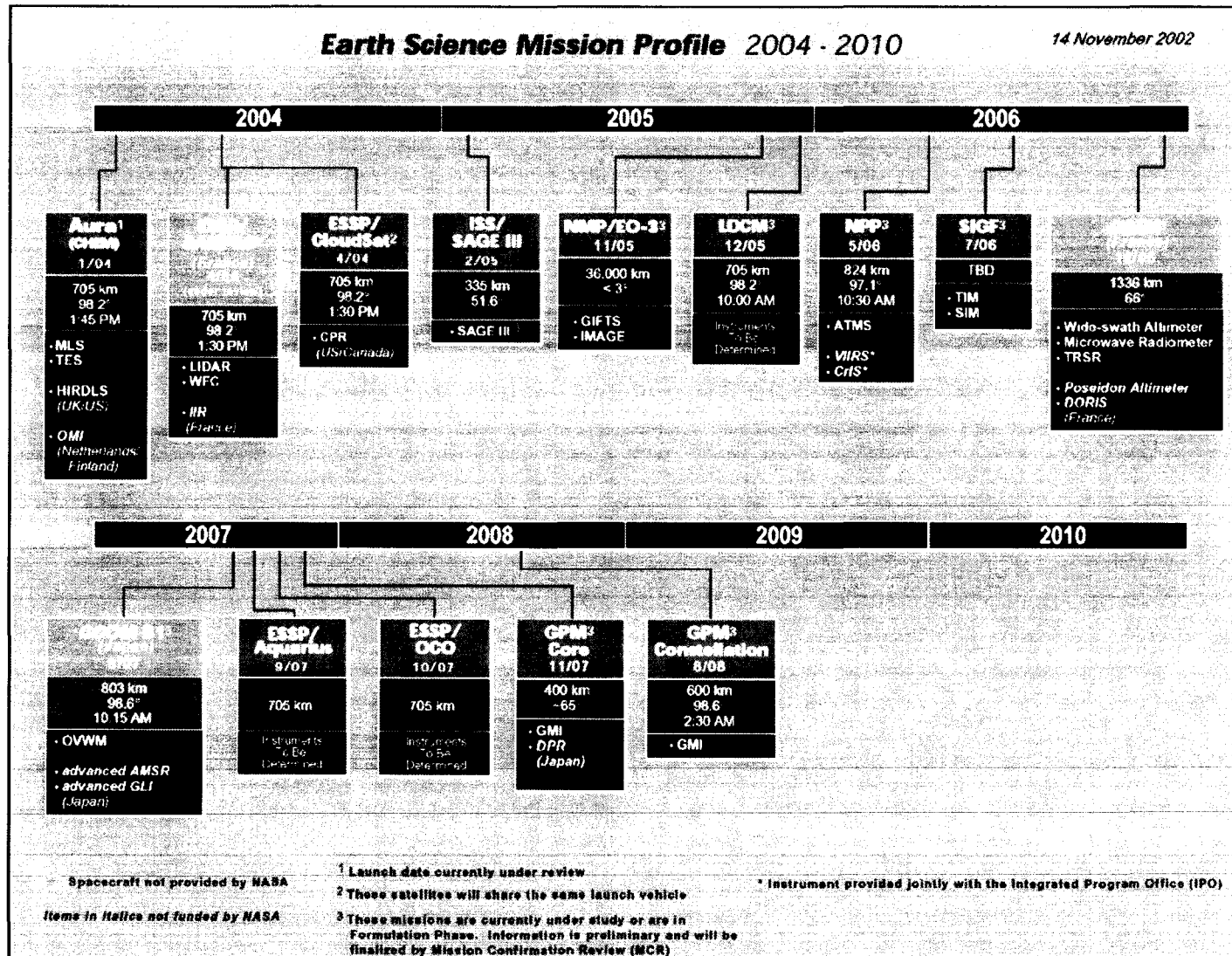


Complex system with many "forcings" and feedbacks:

- atmosphere
- ocean
- land
- biosphere
- cryosphere



# Earth Observing System



# Synergistic Data Sources

- Terra satellite instruments:

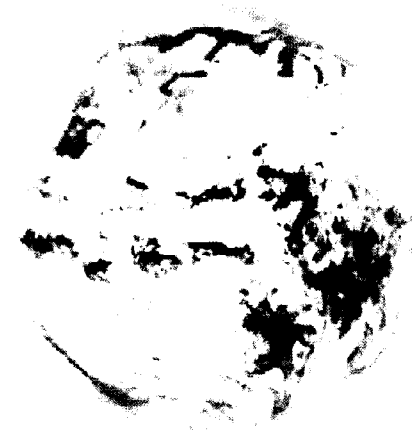
- Advanced Spaceborne Thermal Emission and Reflection Radiometer (ASTER)
- Clouds and the Earth's Radiant Energy Systems (CERES)
- Multi-angle Imaging SpectroRadiometer (MISR)
- Measuring of Pollution in the Troposphere (MOPITT)
- Moderate Resolution Spectrometer (MODIS)

- Aqua satellite instruments:

- Atmospheric Infrared Sounder (AIRS)
- Advance Microwave Radiometer for EOS (AMSR-E)
- Advanced Microwave Sounding Unit (AMSU)
- Clouds and the Earth's Radiant Energy Systems (CERES)
- Humidity Sounder Brazil (HSB)
- Moderate Resolution Spectrometer (MODIS)

- Aura satellite instruments:

- High Resolution Dynamics Limb Sounder (HIRDLS)
- Microwave Limb Sounder (MLS)
- Ozone Measuring Instrument (OMI)
- Tropospheric Emission Spectrometer (TES)



CloudSat

Cloud-Aerosol Lidar and Infrared

Pathfinder Satellite Observation (CALIPSO)

Orbiting Carbon Observatory (OCO)

PARASOL

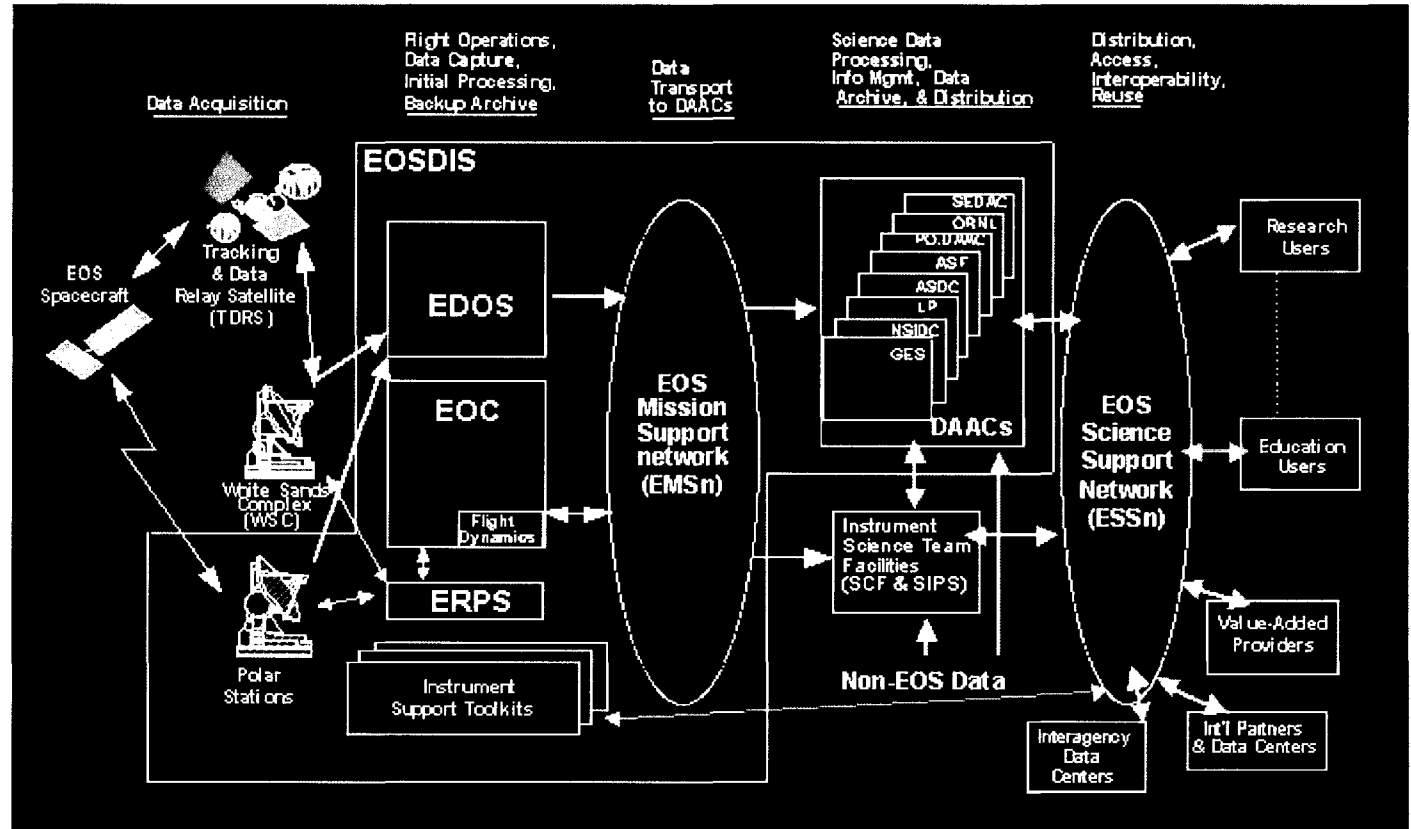
*Technically not part of EOS*



# Data Acquisition

Data downlinked by "granule".

No delay allowed- the firehose is on!





# Data Processing

## Level 0

Reconstructed, unprocessed instrument/payload data at full resolution; any and all communications artifacts, e.g., synchronization frames, communications headers, duplicate data removed.

## Level 1A

Reconstructed, unprocessed instrument data at full resolution, time-referenced, and annotated with ancillary information, including radiometric and geometric calibration coefficients and georeferencing parameters, e.g., platform ephemeris, computed and appended but not applied to the Level 0 data.

## Level 1B

Level 1A data that have been processed to sensor units (not all instruments will have a Level 1B equivalent).

## Level 2

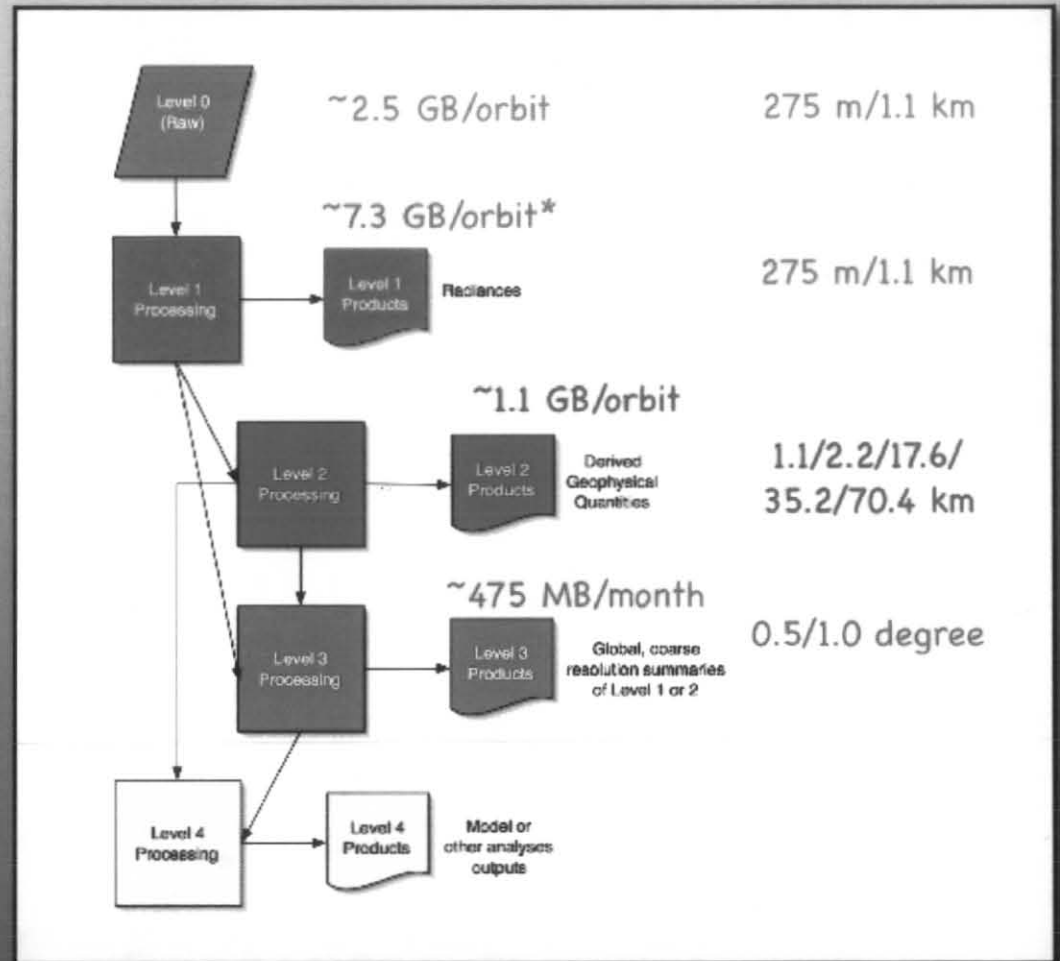
Derived geophysical variables at the same resolution and location as the Level 1 source data.

## Level 3

Variables mapped on uniform space-time grid scales, usually with some completeness and consistency.

## Level 4

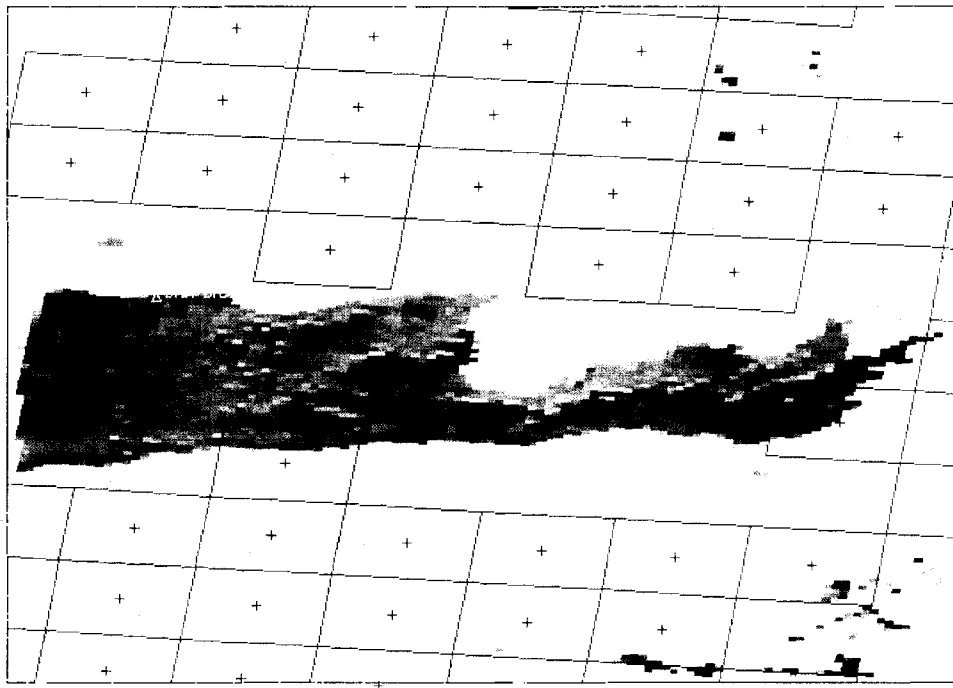
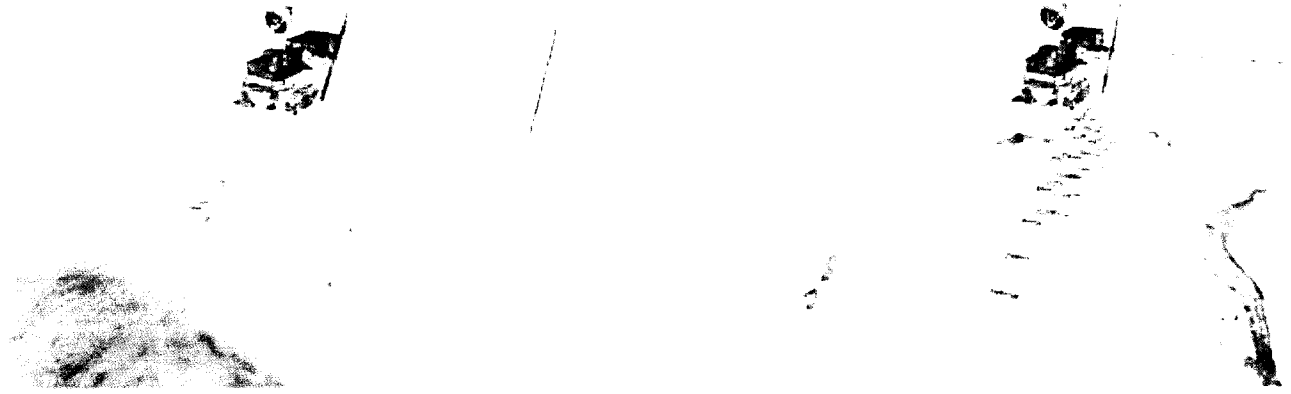
Model output or results from analyses of lower level data, e.g., variables derived from multiple measurements.



# The Challenge

Example:

MISR and MODIS  
on Terra



Sampling patterns of level 2 aerosol data sets; MISR (yellow pixels) and MODIS (blue pixels).

Different resolutions (MISR = 17.6 km, MODIS = 10 km), offset and misaligned grids => irregular overlap pattern.

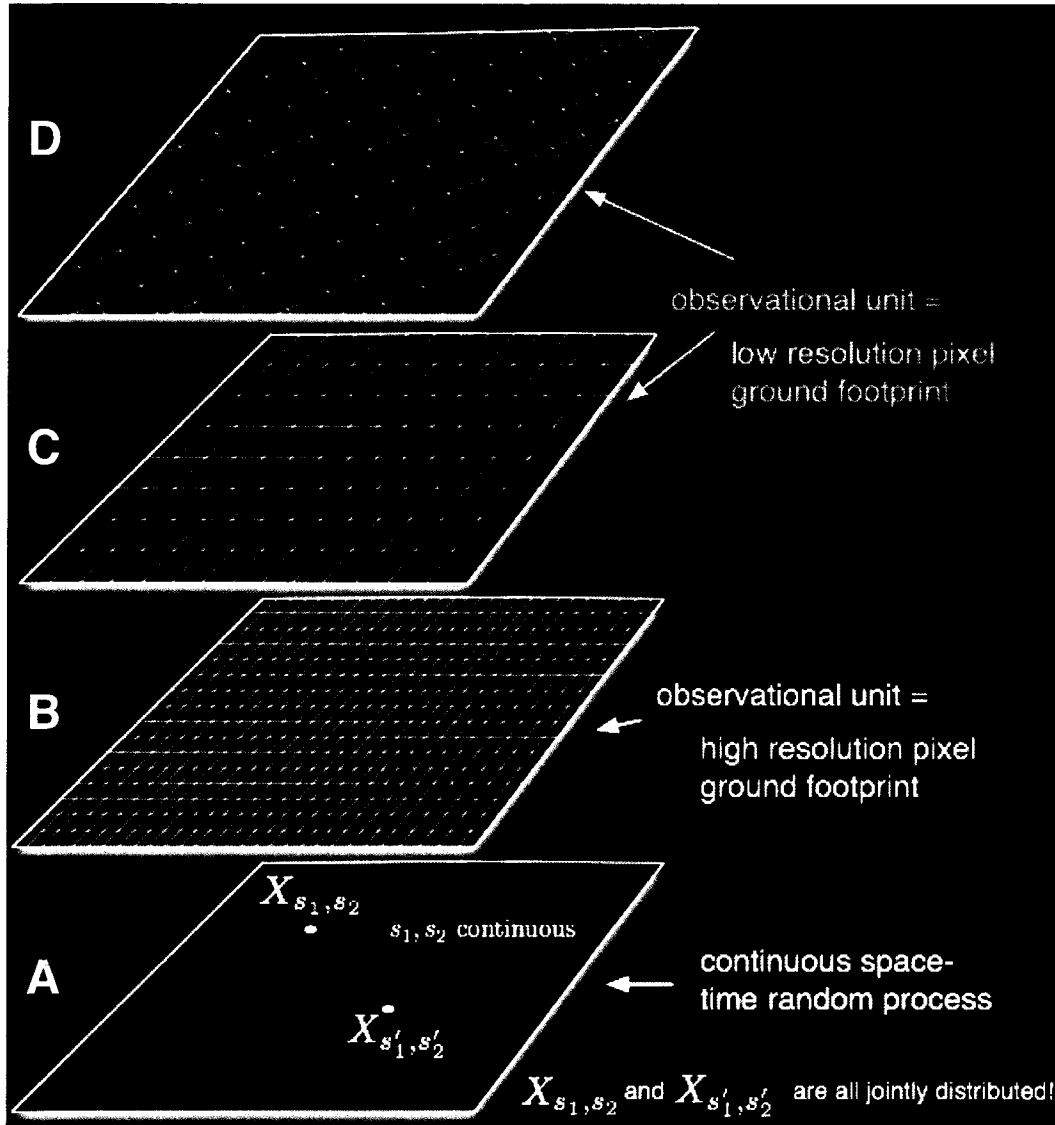




# The Challenge

- What we wish we had:
  - MISR and MODIS measurements on the same, very high resolution grid, at the same time, and with the same measurement error characteristics.
  - We could then estimate the joint distribution of MISR and MODIS empirically, and use it to make inferences about the underlying physical process. (*We still need a statistical model for the physical process to accomplish this.*)
  - Applies just as well whether MISR and MODIS are measuring “the same thing” or “different things”.
- What we do have:
  - MISR and MODIS measurements on different, coarse resolutions (spatial aggregates) overlapping grids at different times\*, and with different measurement error characteristics. (\* In the simple example on the previous slide, we assume same times.)
- Working definition of data fusion: to infer the joint distribution of two quantities measured on different (but related) observational units.

# A Framework



The same measurements taken at different resolutions measure “different” things.

-- thunderstorms vs. hurricanes

Different “windows” on the same underlying phenomena?

-- cloud fraction

Suppose you observe B. How to get to C (nested: easy)? D (non-nested/ misaligned: hard)? A (very hard).

- go B to C by aggregating over observational units
- go to B to A using geostatistical models and methods
- how to go from C (or B) to D? Via A?

# A Spatial Statistics Framework

Statistical model of data generating mechanism at multiple resolutions

Process model:

$$X_s = \mu + \delta_s$$

Data model:

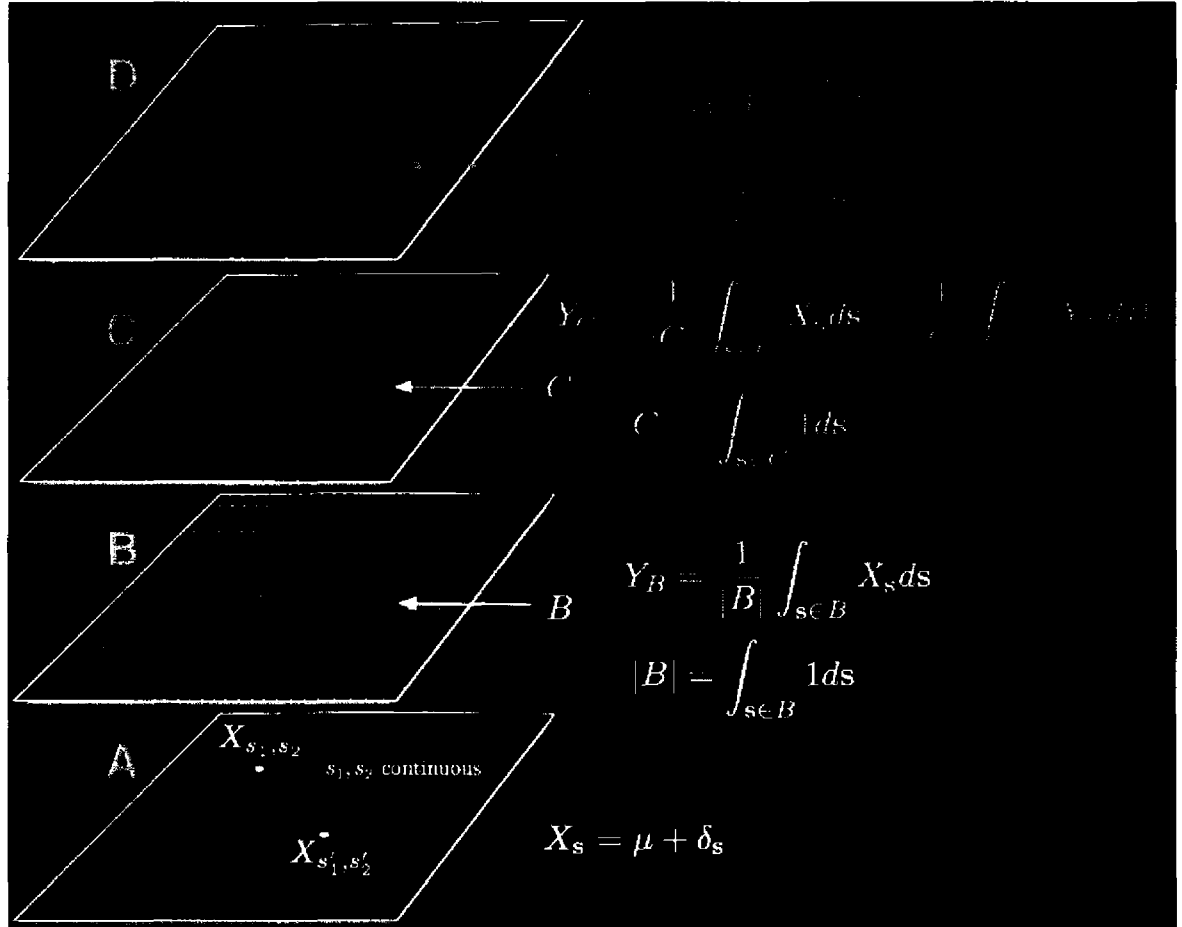
$$Y_B^* = Y_B + \epsilon_B$$

$$= \frac{1}{|B|} \int_{s \in B} (\mu + \delta_s) ds + \epsilon_B$$

$\mu$  = large scale base value

$\delta_s$  = local perturbation

$\epsilon_B$  = measurement error at resolution B

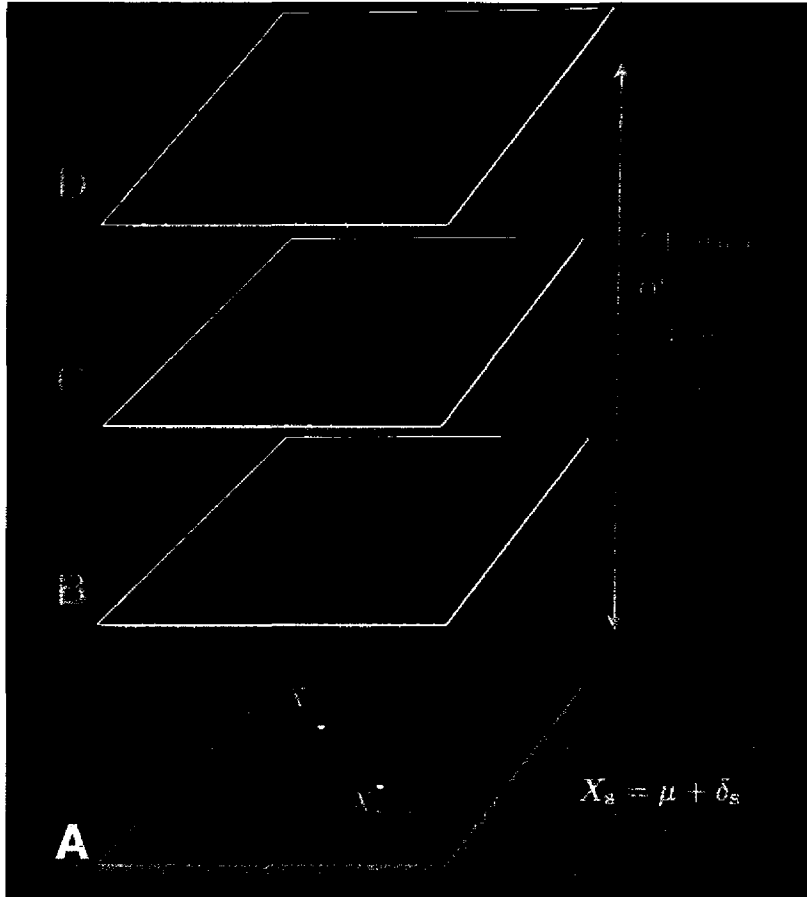


$\mu$  (or  $\mu(s)$ ) may be all or part deterministic

$Cov(\delta_s, \delta_{s'})$  specifies spatial covariance

# A Spatial Statistics Framework

Can also understand relationships between B, C, D by:



1) Estimating spatial covariance by applying geostatistical methods to blocks (pixels):

estimate  $Cov(\delta_s, \delta_{s'})$  from the data as a function of distance and direction and use optimal statistical interpolation to smooth/estimate (kriging).

2) Using a Markov Random Field approach:

specify  $Cov(\delta_s, \delta_{s'})$  as a function of neighborhood structure, i.e. the conditional distribution of  $\delta_s$  given all  $s' \neq s$  depends only on the neighbors of  $s$ .

# A Framework

The objective of data fusion in the MISR/MODIS case is to combine data sets of the forms C (or B) and D but for what purpose?

Does this mean (1) we want to use information provided by C and D to make inferences about the A process...

or

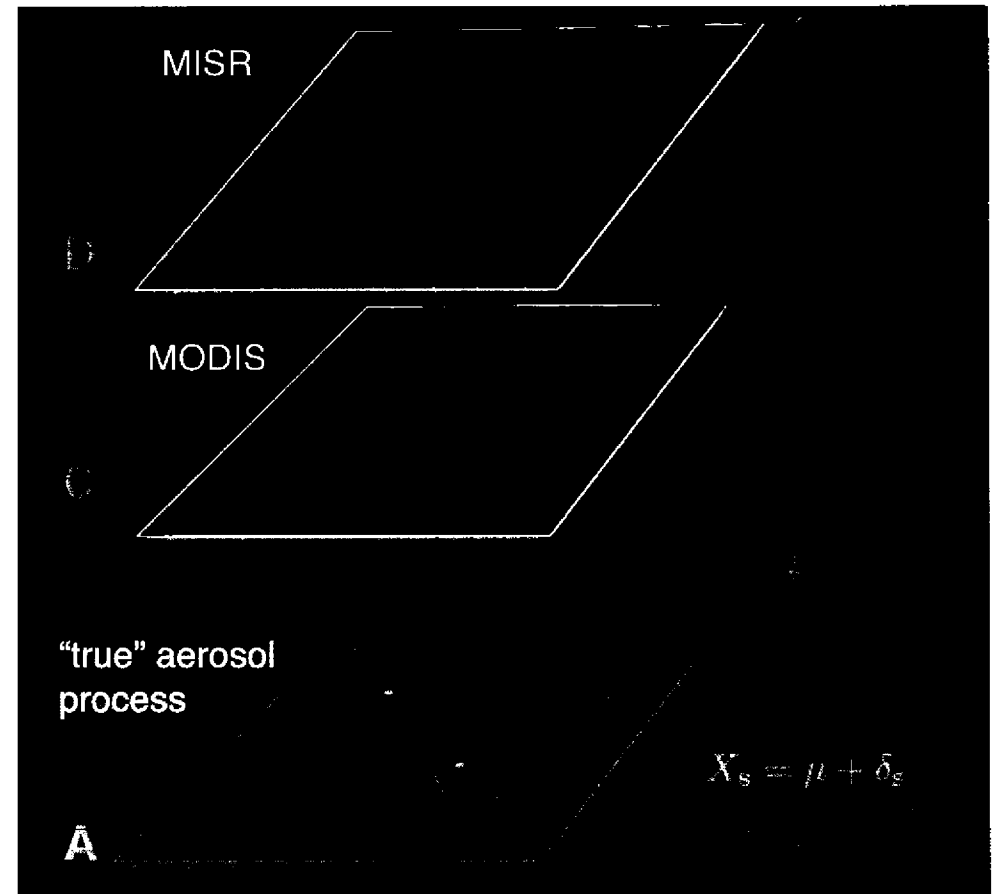
does this mean (2) we want to use information provided by C and D to make inferences about the spatial process on areal units defined by the superposition of D on C?

What are the implications of deciding one way or the other?

*In other words...*

As the size of the observational units in C and D (pixels) gets smaller and smaller (2) approaches (1).

Since both MISR and MODIS measure related quantities we learn more and more about their joint distribution at the level of A as (2) approaches (1). (If they measure the same quantity, we expect their joint distribution to have a linear regression!)





# Conclusions....

- More questions:
  - How does this framework relate to or connect with other frameworks for data fusion? Joint Directors Laboratories “levels” of fusion, for example.
  - Many data fusion methods seek to obtain correct classifications of pixels or objects in images. What are the spatial-statistical models that underlie those methods and how are they related to this problem?
  - Co-registration is not data fusion because co-registration does not address change of support issues. In what settings has co-registration been used as the basis of subsequent analyses such that this distinction matters?
  - What aspects of this spatial statistical framework are transferable and relevant in other domains? Biology? Defense? Medical imaging? Marketing?

End of File

