Outline of Talk

• Motivating problem*
• General Virtual Sensors problem
• Results
• Related accomplishments
• Summary

* Nearly all of this work was presented at the 2005 AGU conference.
Motivating Problem

Many remote sensing problems data analysis problems can be broken down into two components:

1. Seasonal variation
2. Variation induced by the model class

The purpose of this study is to develop algorithms that help us identify these variations separately.

It is necessary to understand this variation in order to develop stable models that can predict energy in one spectral band based on the energy in other spectral bands.

We address these problems through the development of a Virtual Sensor.
Emulating Sensor Signals back in Time for Cloud Trending

No Black Signal for AVHRR/2. Need to emulate AVHRR/2 cloud signal

MODIS 1.6 micron Channel 6 (Black)
Cloud Detection back in Time

Solution: Predict 1.6\(\mu\)m channel using a Virtual Sensor

MODIS 1.6 micron Signal (Black)

Black Signal for AVHRR/2 estimated by a Virtual Sensor.
The Virtual Sensors concept can also be used to deal with multi-resolution analysis:

MODIS Channels 1 & 2: 250 m resolution
MODIS Channel 6: 500 m resolution.

Note: Channel 6 (at 1.6 microns) is not available at the 250 m resolution.
Given:
- MODIS channels 1, 2, 20, 31, 32 correspond to five AVHRR/2 channels

Develop:
- Model MODIS channel 6 (1.6μm) as a function of five MODIS channels

Apply:
- Use function to construct estimate of 1.6μm channel for AVHRR/2
Virtual Sensors

Virtual Sensors predict the historical record for spectral measurements using relationships found from existing sensors and inputs from historical record.

Useful for simulating sensors back in time or multi resolution analyses.

Accuracy of learned models for MODIS data: 70%-90% (over 2 weeks)
Model Classes

Linear Models:
Least squares regression (used as a very simple baseline)

Nonlinear Models:
Neural nets (used as a simple baseline)
Gaussian Processes & Kernel Methods

Data in original space: highly complex decision boundaries.

Data in high dimensional feature space can yield simple decision boundaries.
Linear Correlation Matrix for MODIS Channels over Fresno CA in 2005
Correlation Matrices Changes with location
Seasonal Variation for 5 input channels

Normalized Error for Virtual Sensor built with 5 channels and 1 day of training data

Normalized Error for Virtual Sensor built with 5 channels and 3 days of training data
Seasonal Variation for 2 input channels

Normalized Error for Virtual Sensor built with 2 channels and 1 day of training data

Normalized Error for Virtual Sensor built with 2 channels and 3 days of training data
Seasonal Variation for 2 input channels

Normalized Error for Virtual Sensor built with 2 channels and 1 day of training data

Normalized Error for Virtual Sensor built with 2 channels and 3 days of training data

Black Diamonds: Linear Model
Multi-Resolution Predictions

500 m Image

250 m GP

250 m Bagged Neural Nets

250 m Linear Model
Distributions of Multiresolution Predictions

PDF of Ch. 1 at 500 m Resolution

PDF of Ch. 1 at 250 m Resolution

PDF of Ch. 2 at 500 m Resolution

PDF of Ch. 2 at 250 m Resolution

PDF of Ch. 6 at 500 m Resolution

GP PDF of Ch. 6 at 250 m Resolution

NN PDF of Ch. 6 at 250 m Resolution

Lin PDF of Ch. 6 at 250 m Resolution
The model class significantly affects the model’s stability:

Linear Models: Produce higher overall error and are less robust to seasonal variation.

Nonlinear Models: Produce lower overall error and are more robust with respect to seasonal variations.

These results support the idea that a Virtual Sensor can be used to characterize sensor measurements through time or at different resolutions.

Can be used to reduce processing times significantly for some applications.