

Retrieval of Land Surface Parameters from MODIS and MISR Albedo Products

B. Pinty⁽¹⁾, M. Clerici⁽¹⁾

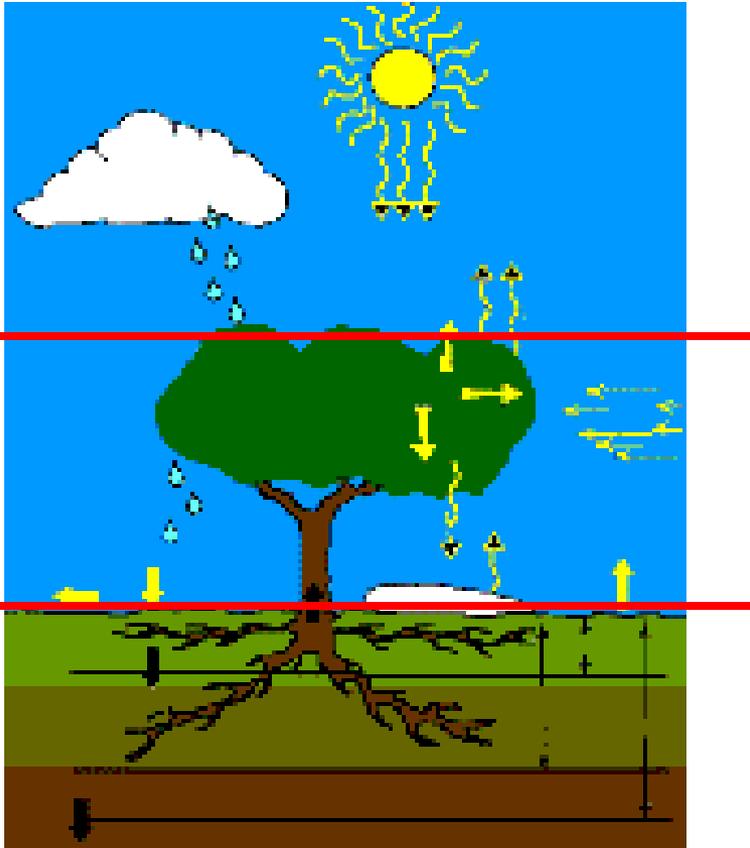
I. Andredakis⁽¹⁾, M. Vossbeck⁽³⁾ T. Lavergne⁽²⁾,
T. Kaminski⁽³⁾ M. Taberner⁽¹⁾, N. Gobron⁽¹⁾

1) EC-JRC, IES, Ispra, Italy

2) Norwegian Meteorological Institute, Blindern, Oslo

3) FastOpt, Hamburg, Germany

Energy partitioning between the vegetation and the soil layer



- The **surface albedo** corresponds to the upper boundary condition of the vegetation plus soil RT and other problems
Need to understand and represent the RT processes yielding the distribution of energy below that “surface”, e.g., **transmitted fluxes**.
- The remaining energy in the soil “layer” is used to solve the heat conduction equation and soil hydrology, e.g., snow melting, evaporation.
- The energy left into the vegetation “layer” is used to drive the water, e.g., evapotranspiration, and the carbon cycle, e.g., NPP, NEP,..

Challenges for the Land community

- 4 identified ECVs namely **Albedo**, **FAPAR**, **LAI** and ultimately **Land Cover** are linked via radiation and phenological processes.
- **Multiple datasets** of these ECVs are available from different institutions.

The retrieval of these ECVs can be formulated in order to find solutions **optimizing all the available information** i.e., inferring statistically the state of the system

Challenges for the Land community

- 4 identified ECVs namely **Albedo**, **FAPAR**, **LAI** and ultimately **Land Cover** are linked via radiation and phenological processes:

They **MUST** be **retrieved consistently** and then specified in the same manner in host models

- **Multiple datasets** of these ECVs are available from different institutions:

They **MUST** be analyzed and exploited to establish a **coherent set of information** across platforms and institutions.

Retrievals of model Parameters for Land surface schemes

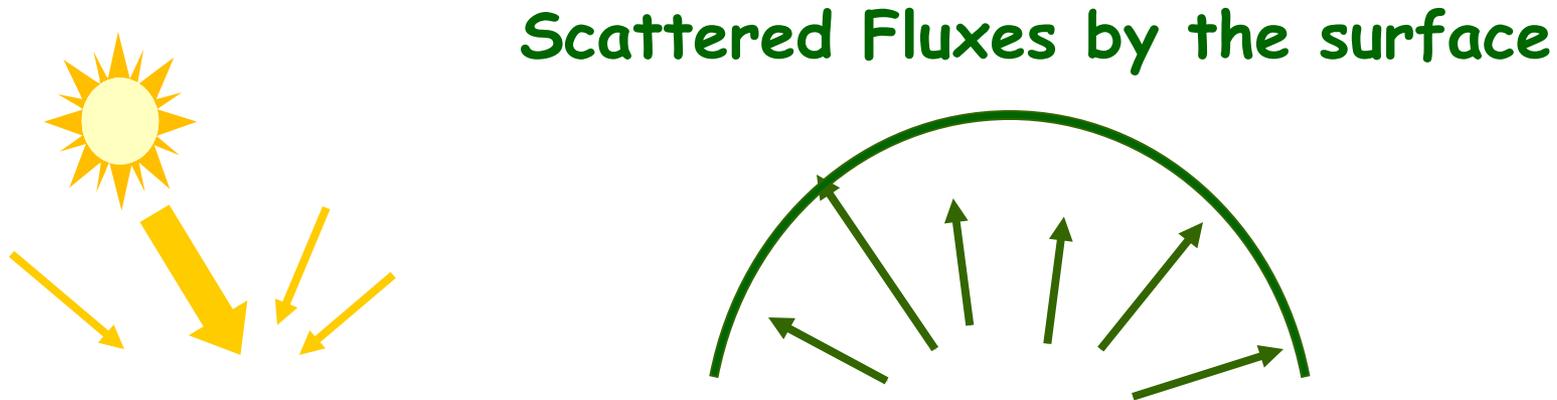
The inverse problem must be formulated in order to find solutions **optimizing all the available information** i.e., inferring statistically the state of the system

Towards an **integrated system** for the optimal use of remote sensing flux products

Towards an integrated system for the optimal use of remote sensing products

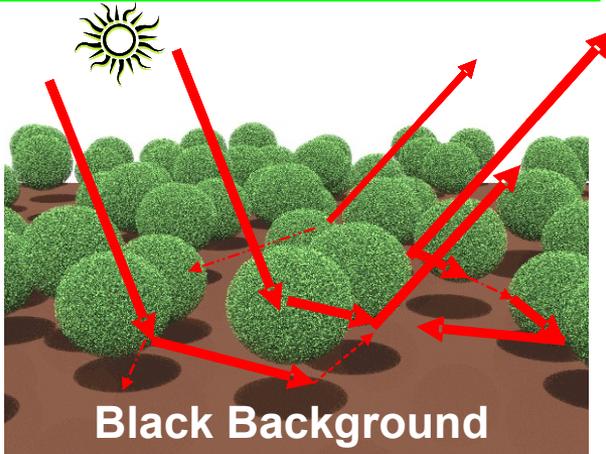
- 1- Develop a **RT flux model** to be used in forward/assimilation mode in SVAT modules of large scale models
- 2- Infer the state variables of the RT flux model from satellite products, e.g., **surface albedo**

Two-stream model to redistribute Sun energy between the atmosphere and the biosphere



Design of the two-stream flux model

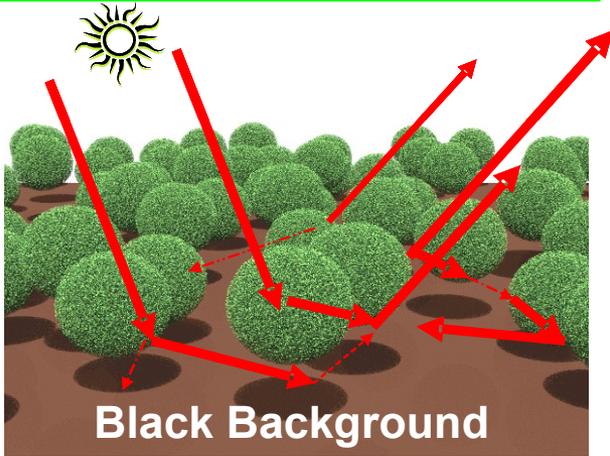
Amount of leaf material and leaf color



- Regulates the *absorption* processes associated with vegetation photosynthesis

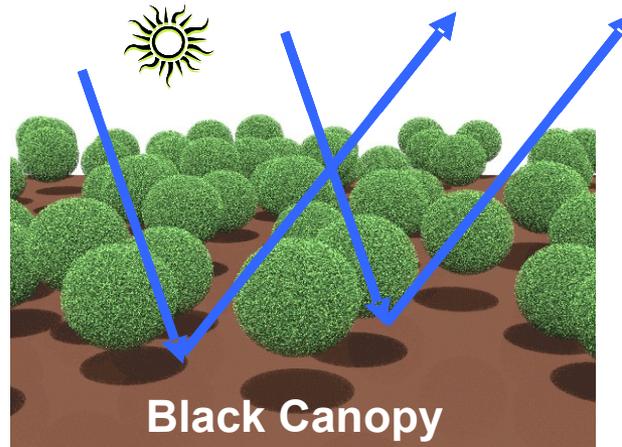
Design of the two-stream flux model

Amount of leaf material and leaf color



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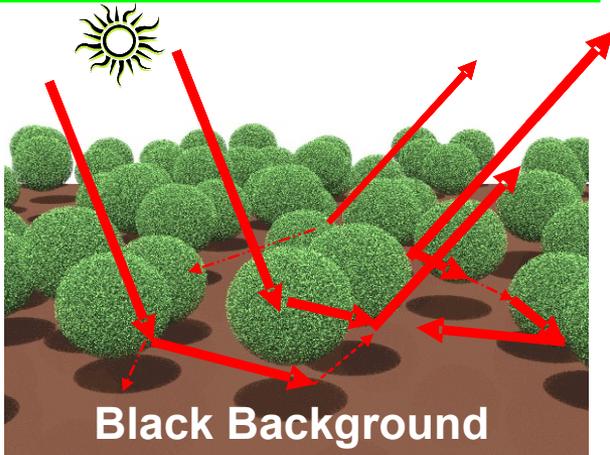
Distribution of leaves and background albedo



- No absorption process by vegetation associated with this wavelength-independent contribution

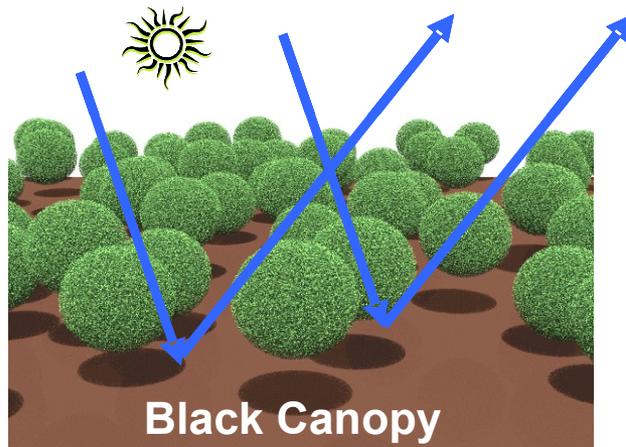
Design of the two-stream flux model

Amount of leaf material and leaf color

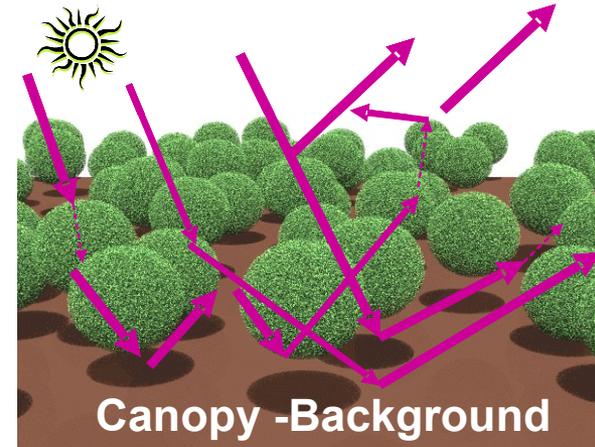


- Regulates the *absorption* processes associated with vegetation photosynthesis

Distribution of leaves and background albedo



- No absorption process by vegetation associated with this wavelength-independent contribution



- Controlled by multiple scattering events between the background and the canopy

Requirements from 2-stream model

- 3 (effective) state variables:

1. *Optical depth: LAI* *amount of leaf material*

2. *single scattering albedo :*

Leaf reflectance+ Leaf transmittance *leaf color*

3. *asymmetry of the phase function*

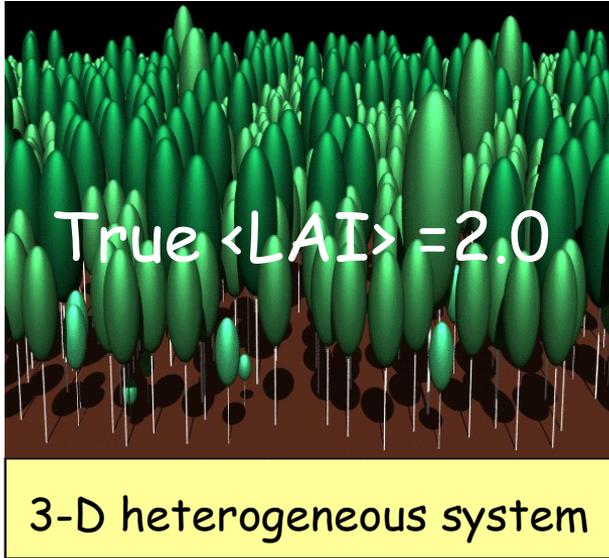
Leaf reflectance/transmittance

- 2 boundary conditions:

1. *Top: Direct and Diffuse atmospheric fluxes (known)*

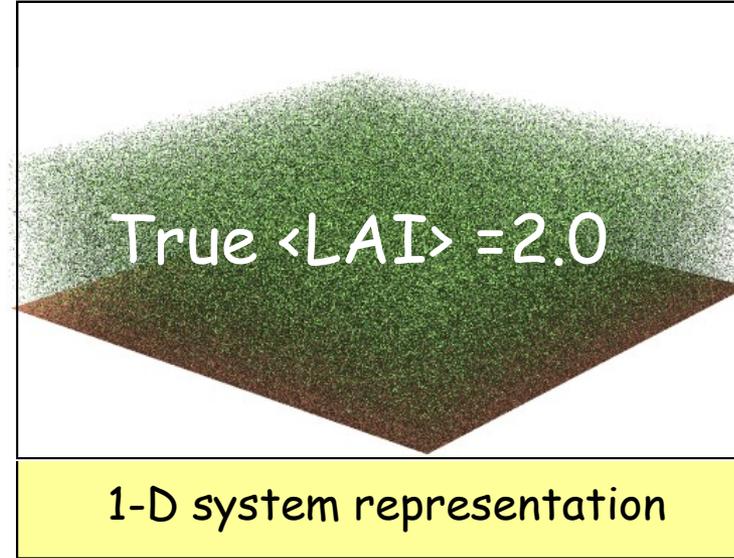
2. *Bottom : Flux from **background Albedo** (unknown) *soil color**

The concept of effective LAI



Direct transmission at 30 degrees Sun zenith angle,

$$T_{3-D}^{direct}(\langle LAI \rangle) = 0.596$$



Direct transmission at 30 degrees Sun zenith angle,

$$T_{1-D}^{direct}(\langle LAI \rangle) = \exp\left(-\frac{\langle LAI \rangle}{2\mu_0}\right) = 0.312$$

Effects induced by internal variability of LAI

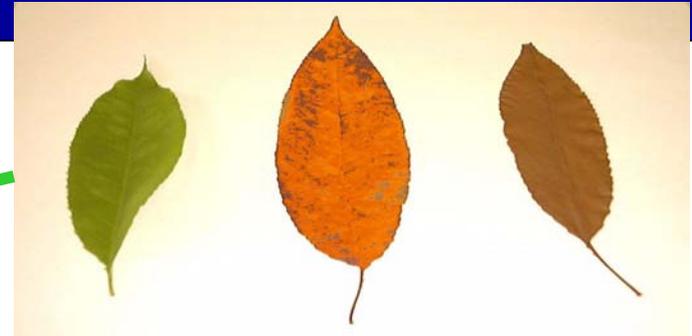
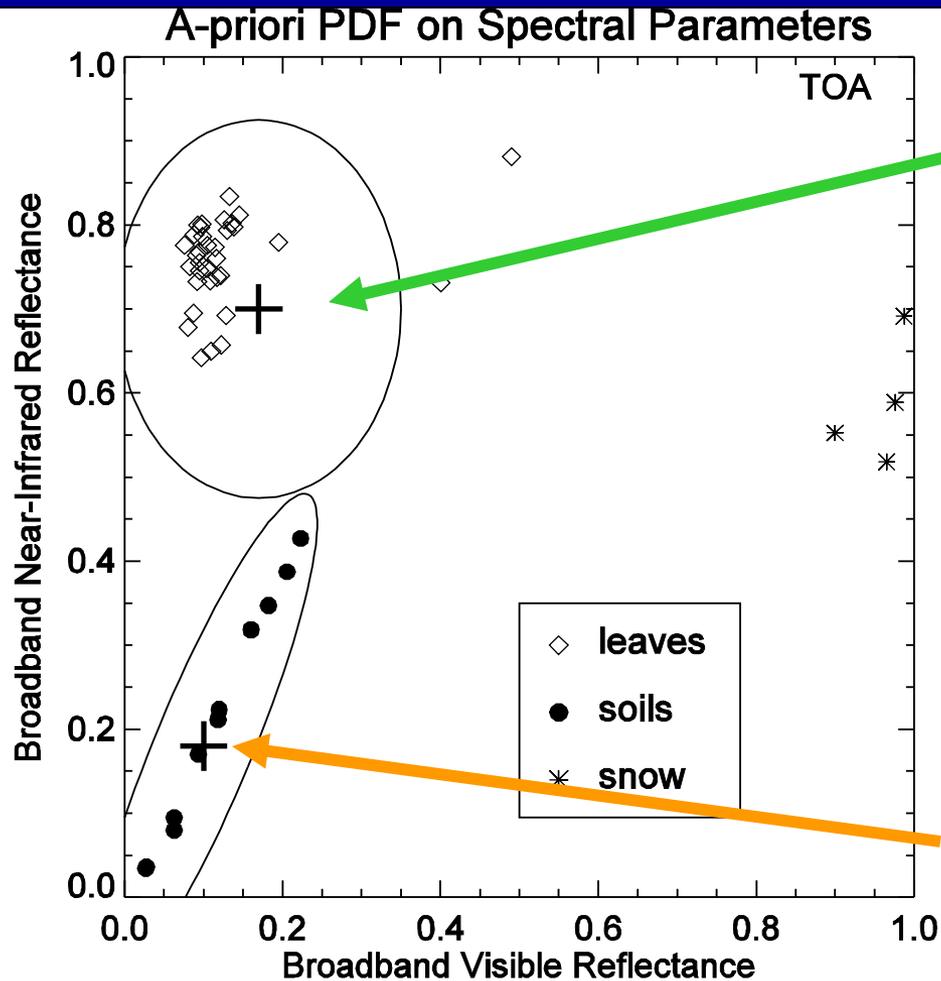
INPUTS : observations/model

- Remote Sensing Flux products, e.g. **Albedo**
Vis/NIR, noted **d**
- Updated/benchmarked **2-stream model** from
Pinty
et al., noted $M(\mathbf{X})$

INPUTS : observations/model

- Remote Sensing Flux products, e.g. **Albedo Vis/NIR**, noted \mathbf{d}
- **uncertainty** on the RS products is specified in the measurement set covariance matrix \mathbf{C}_d
- A priori knowledge/guess on **model parameters** noted \mathbf{X}_{prior}
- **uncertainty** associated the model parameter is specified via a covariance matrix $\mathbf{C}_{X_{prior}}$

prior knowledge on model parameters

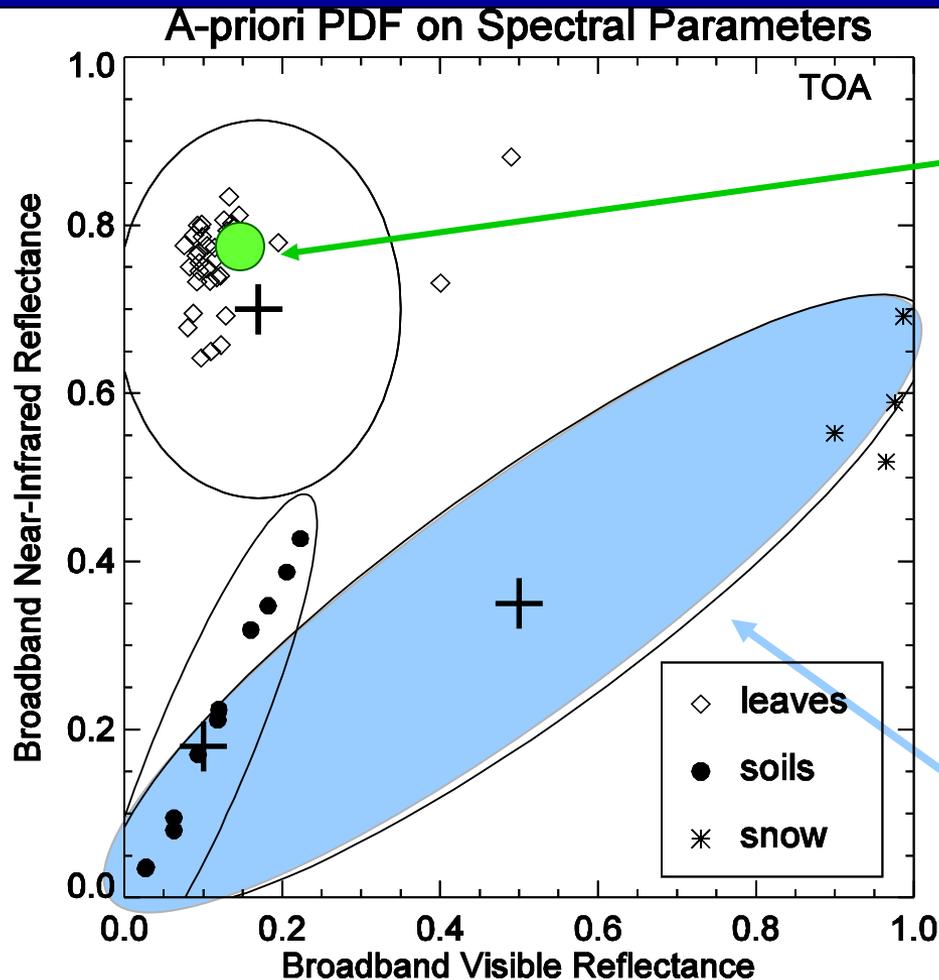


$$LAI_{prior} = 1.5$$

$$\sigma_{prior}(LAI) = 5.0$$



prior knowledge on model parameters



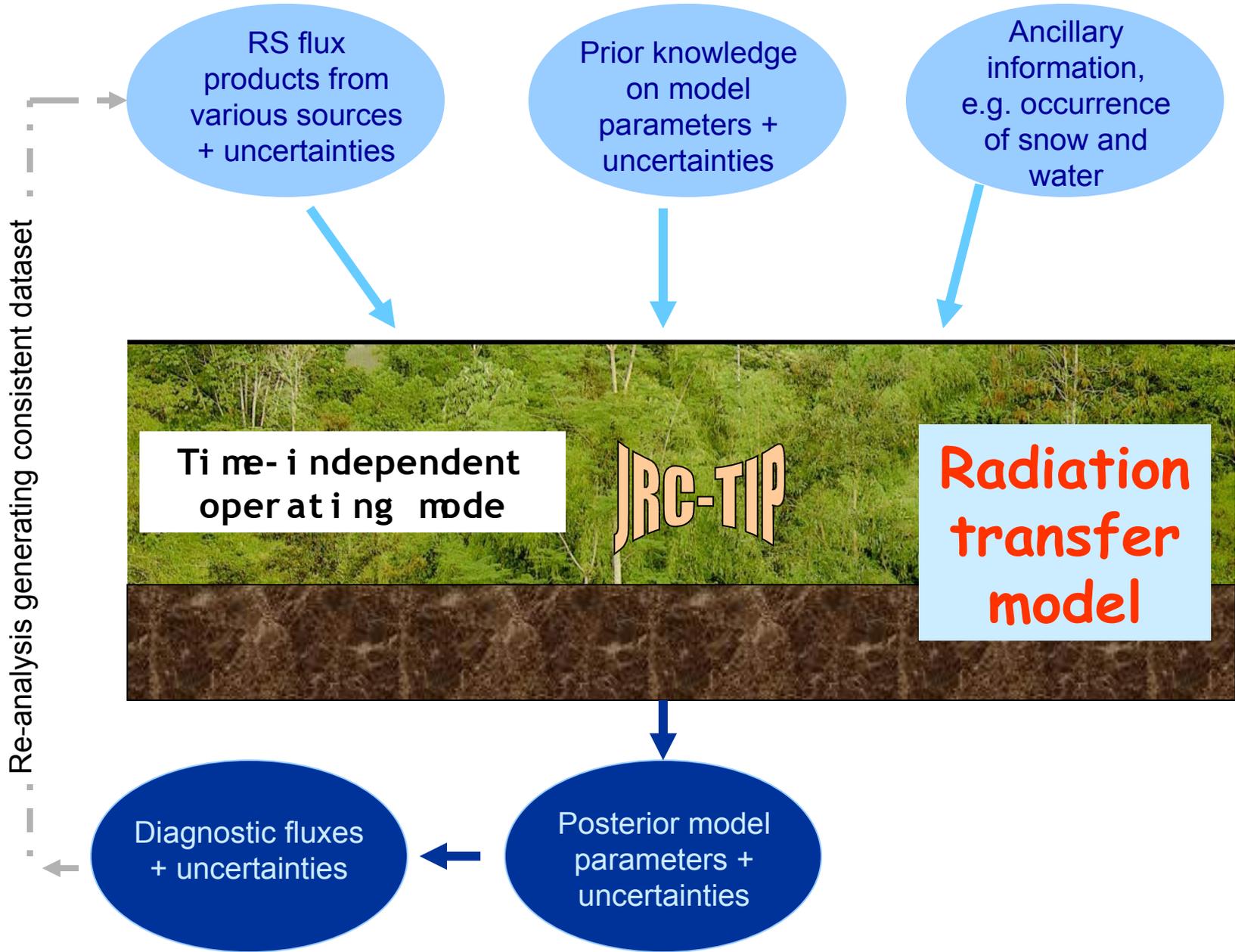
a priori 'green' leaves

$$LAI_{prior} = 1.5$$

$$\sigma_{prior}(LAI) = 5.0$$

in case snow occurs

JRC-Two-stream Inversion Package: JRC-TIP



The core of the JRC-TIP

$$J(\mathbf{X}) = \frac{1}{2} \left[(M(\mathbf{X}) - \mathbf{d})^T \mathbf{C}_d^{-1} (M(\mathbf{X}) - \mathbf{d}) + (\mathbf{X} - \mathbf{X}_{prior})^T \mathbf{C}_{X_{prior}}^{-1} (\mathbf{X} - \mathbf{X}_{prior}) \right]$$

Model parameters

2-stream model

measurements

Parameter knowledge

Uncertainty measurements

Uncertainty parameters

- Computer optimized **Adjoint** and **Hessian model** of cost function from automatic differentiation technique (assume Gaussian theory).
- PDFs of **all** 2-stream model parameters
- Assessment of **all fluxes** predicted by the 2-stream model and their associated uncertainty

OUTPUTS: posterior knowledge

- PDFs of **all** 2-stream model parameters:

$$PDF(\mathbf{X}) \approx \exp\left(-\frac{1}{2}(\mathbf{X} - \mathbf{X}_{post})^T \mathbf{C}_{X_{post}}^{-1} (\mathbf{X} - \mathbf{X}_{post})\right)$$

a posteriori uncertainty covariance matrix

- Assessment of **all fluxes** predicted by the 2-stream model and their associated uncertainty:

$$\mathbf{C}_{post}^{Flux} = \mathbf{G} \mathbf{C}_{X_{post}} \mathbf{G}^T$$

Operational Processing of MODIS – MISR albedo products

JRC-TIP applied to global scale, 1 Km resolution yearly time-series (2005) of MODIS and MISR albedos (BHRs).

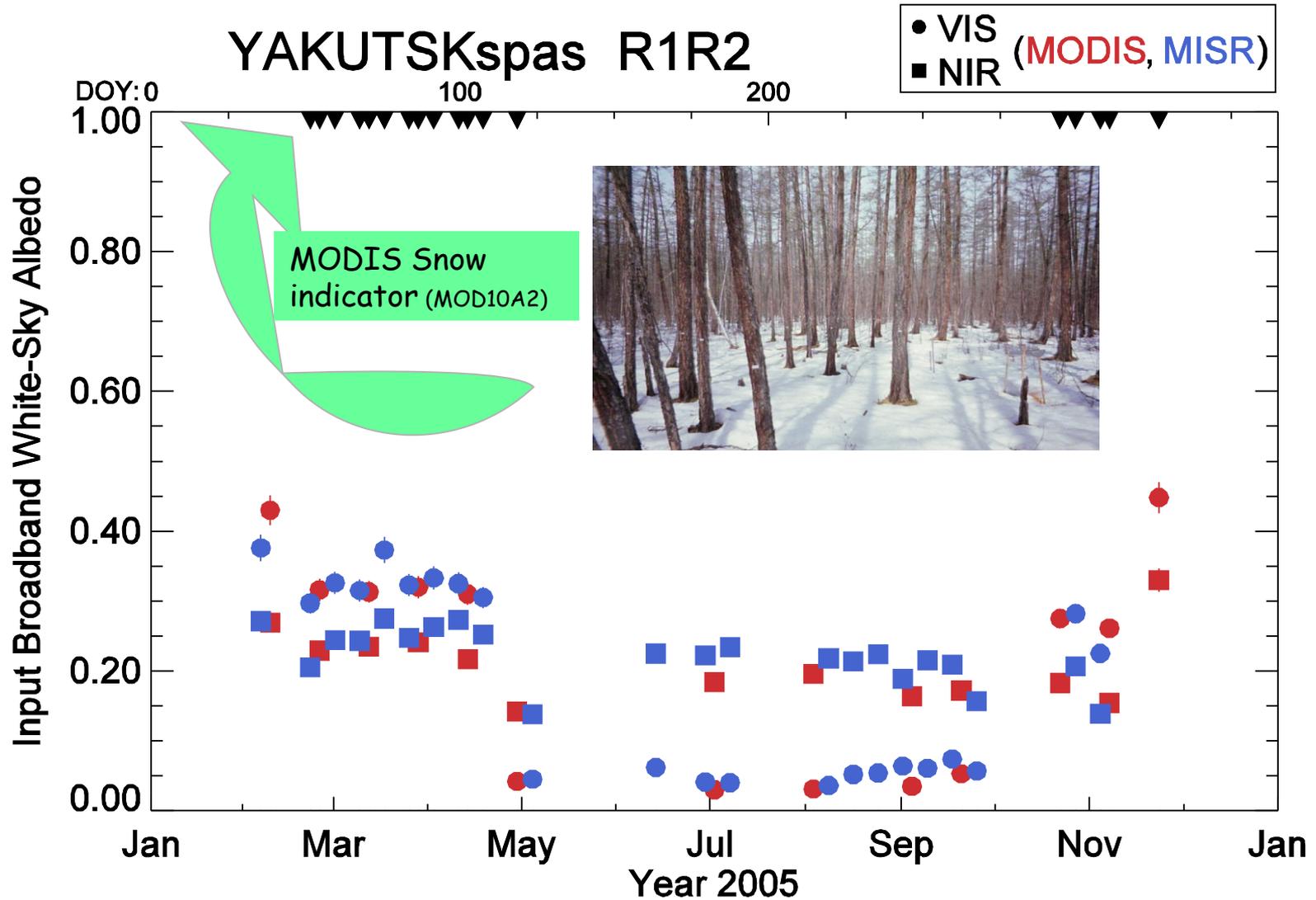
	Total Number of Inversions	Valid Results (%)	Solution not Found (%)	High Residual misfit (%)
MODIS	1.88×10^8	98.1	1.84	0.11
MISR	2.72×10^8	93.8	4.24	1.99

Application over Yakustk Forest: a deciduous needle-leaf larch forest



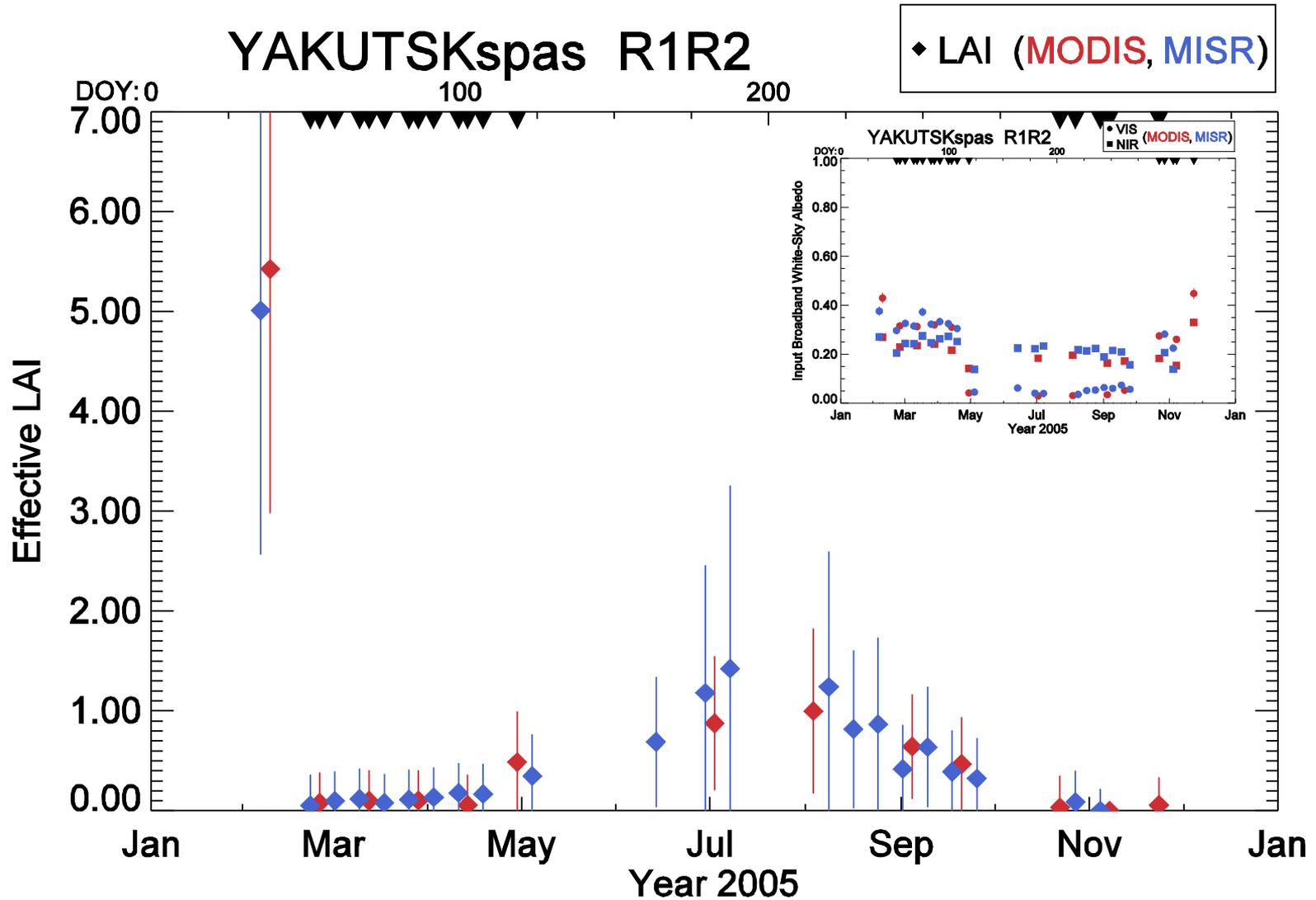
Courtesy of Dr. R. Suzuki

Application over Yakutsk: Measurements

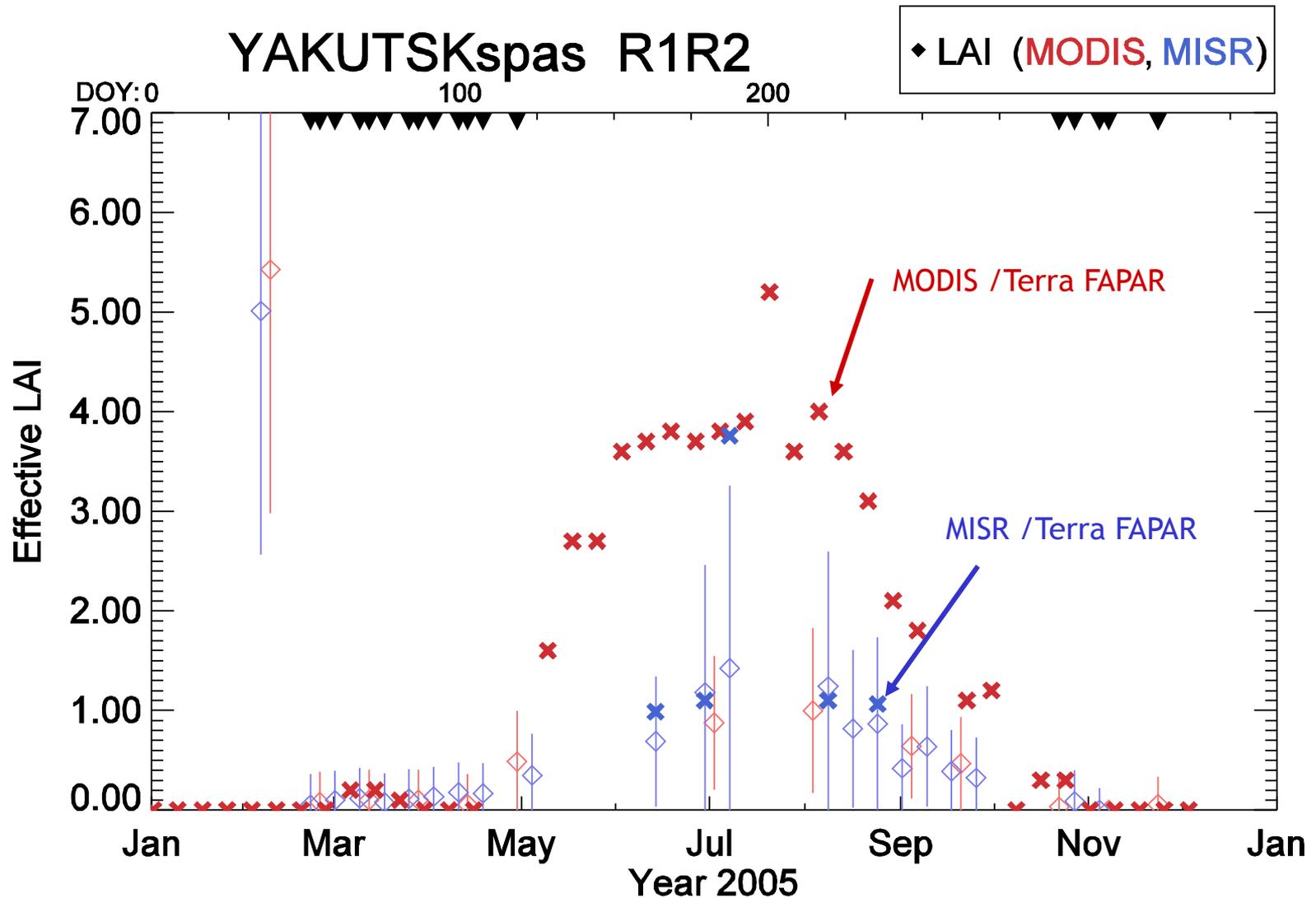


Specified uncertainty on BHRs is 5% relative

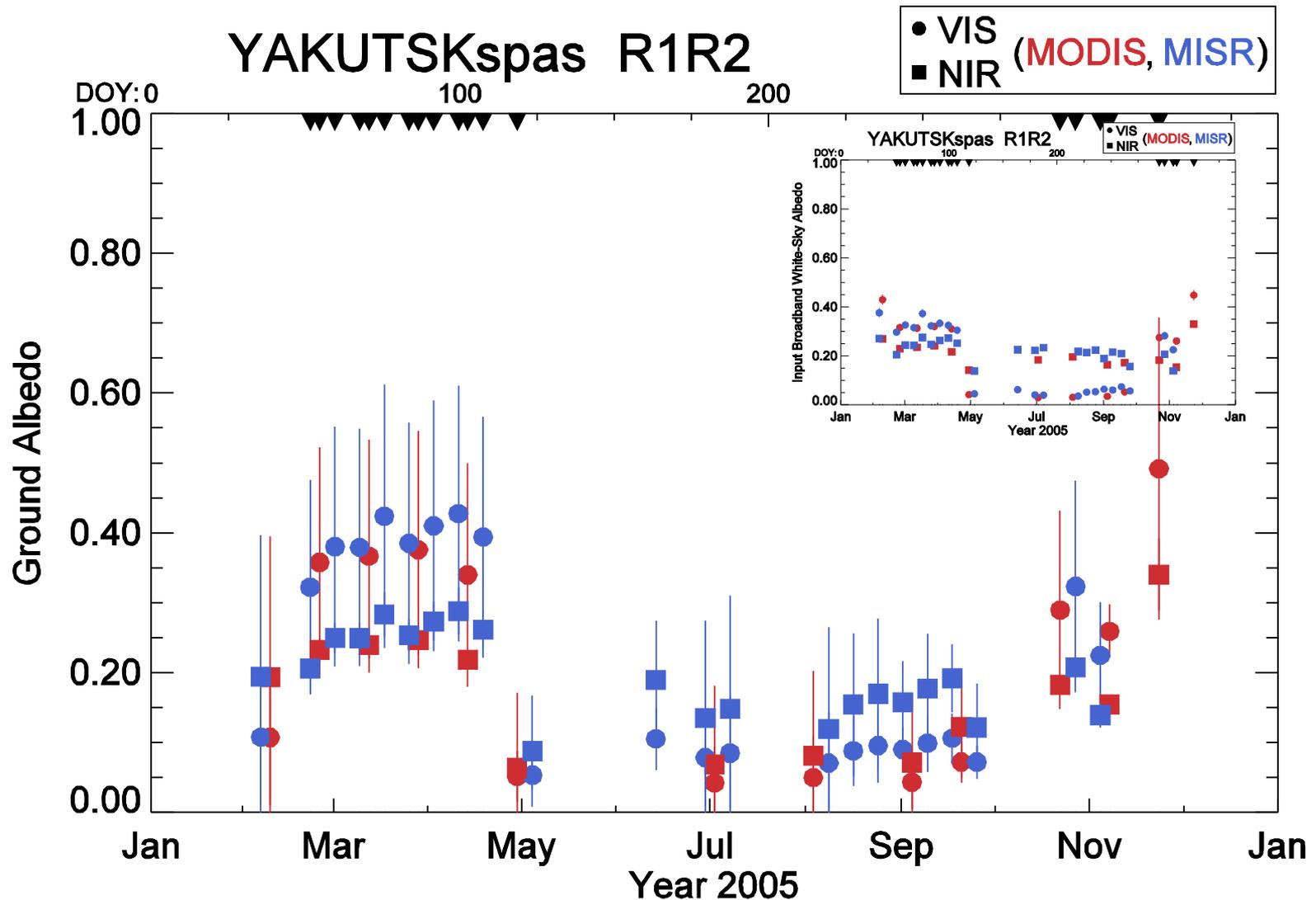
Application over Yakutsk: model parameters



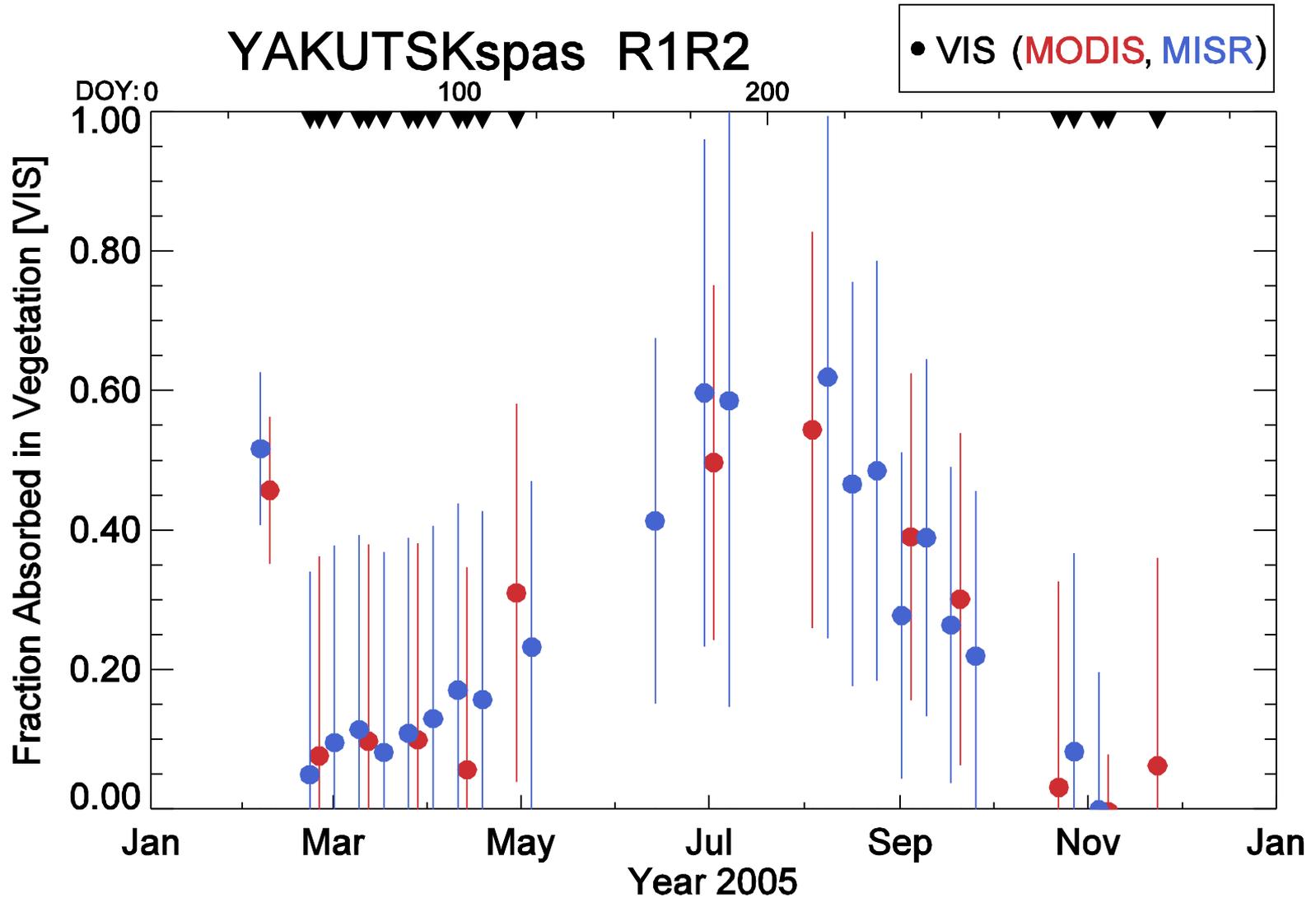
Application over Yakutsk: model parameters



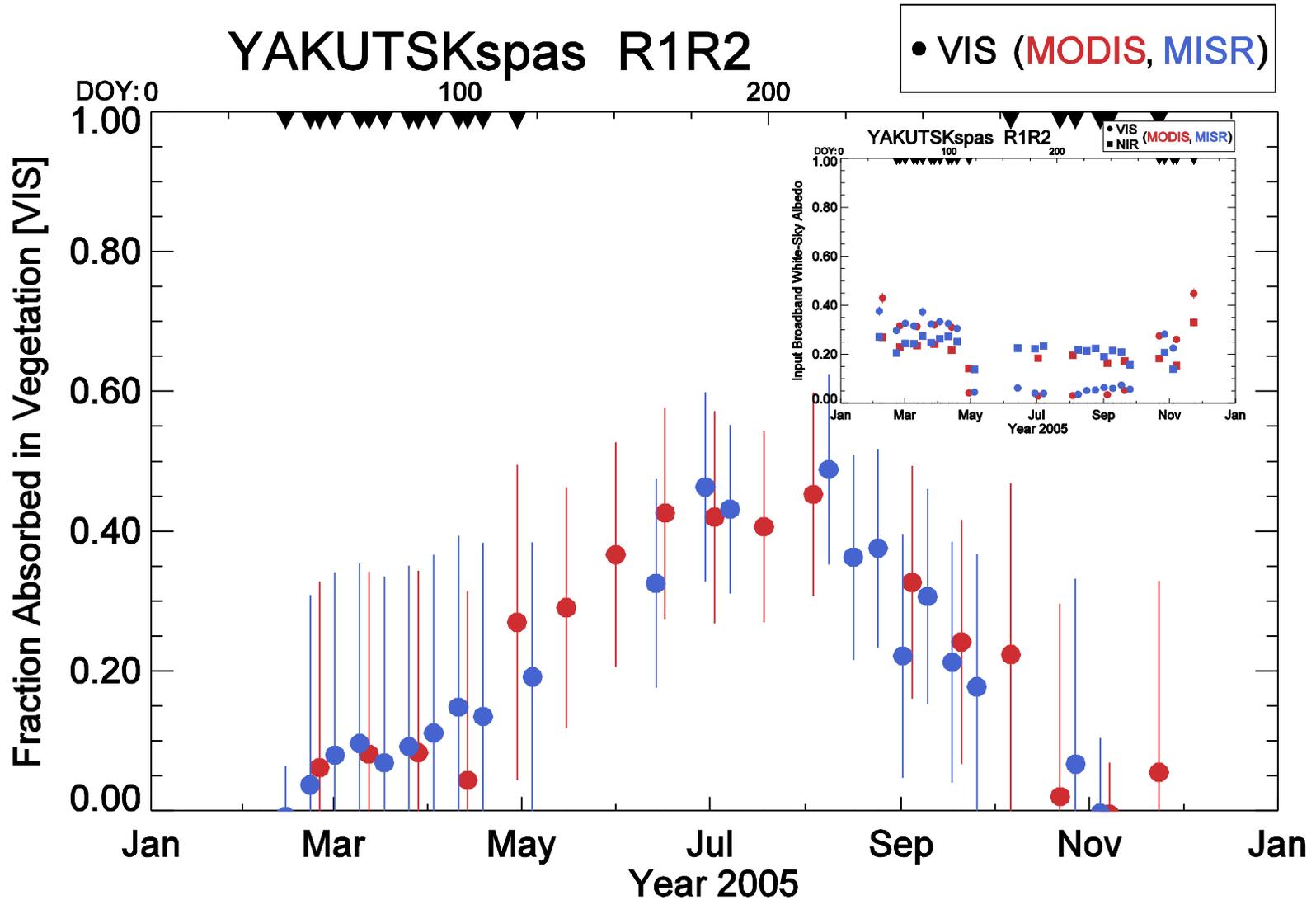
Application over Yakutsk: model parameters



Application over Yakutsk: radiant fluxes

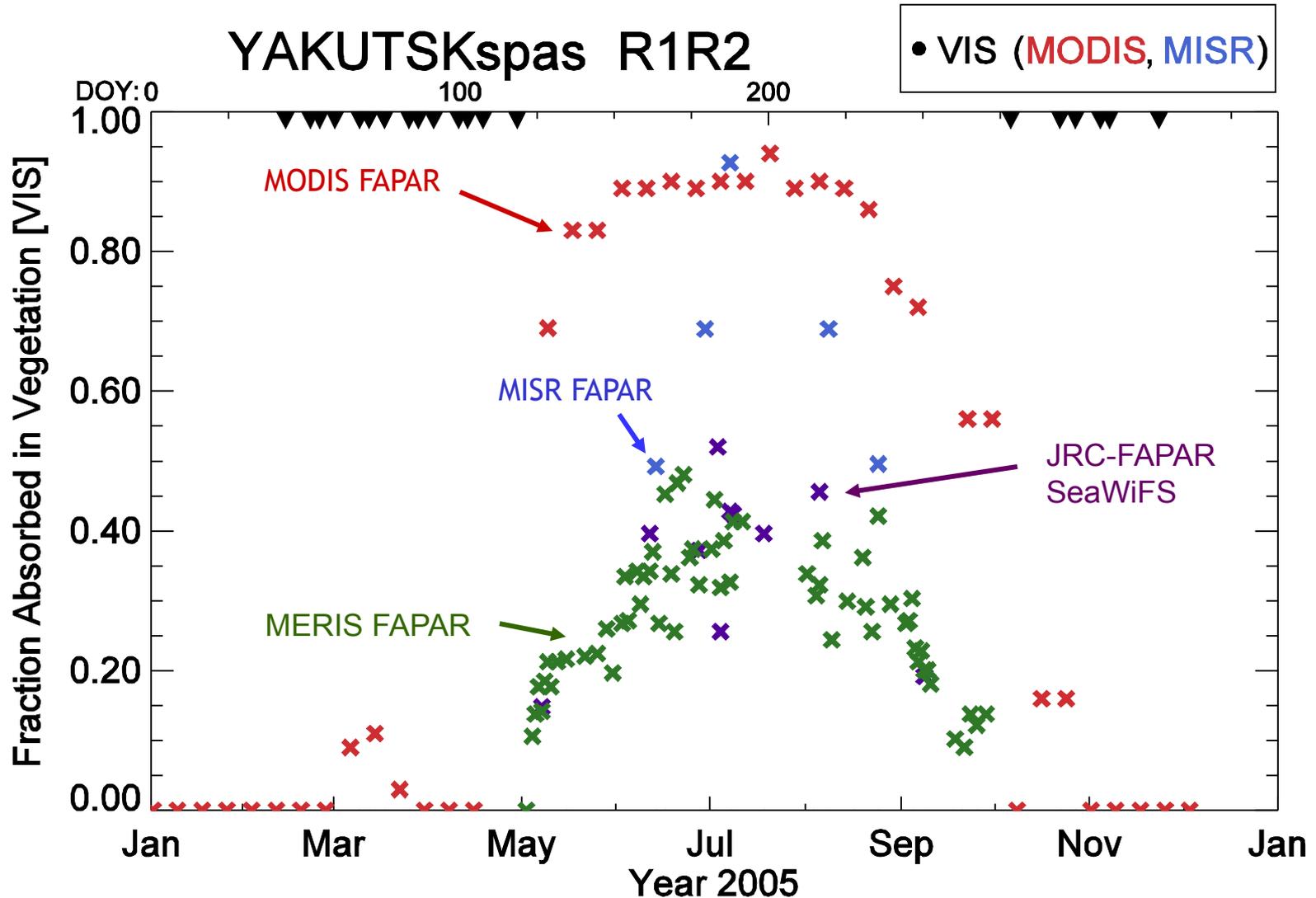


Application over Yakutsk: radiant fluxes

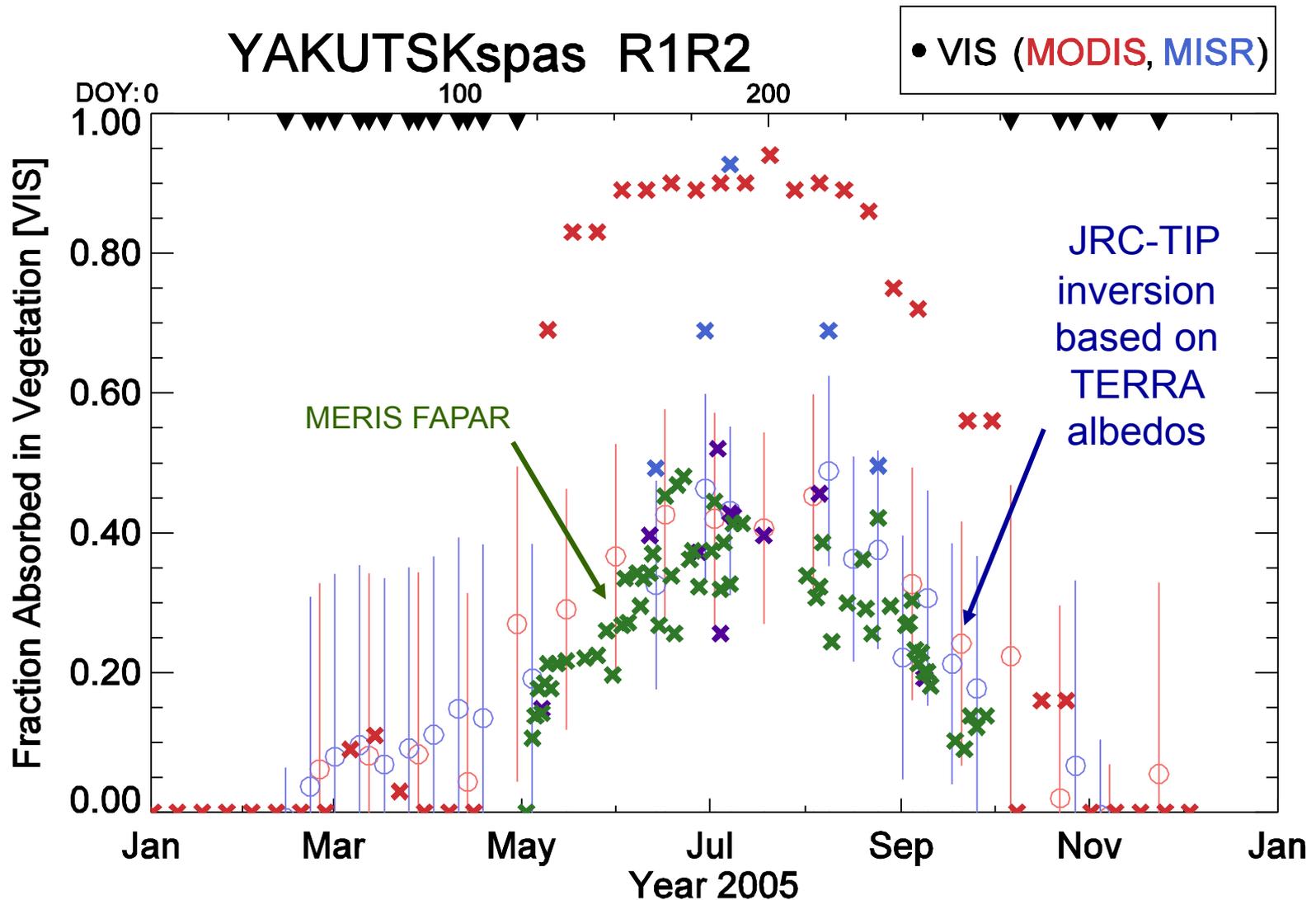


a priori 'green' leaves

Application over Yakutsk: radiant fluxes



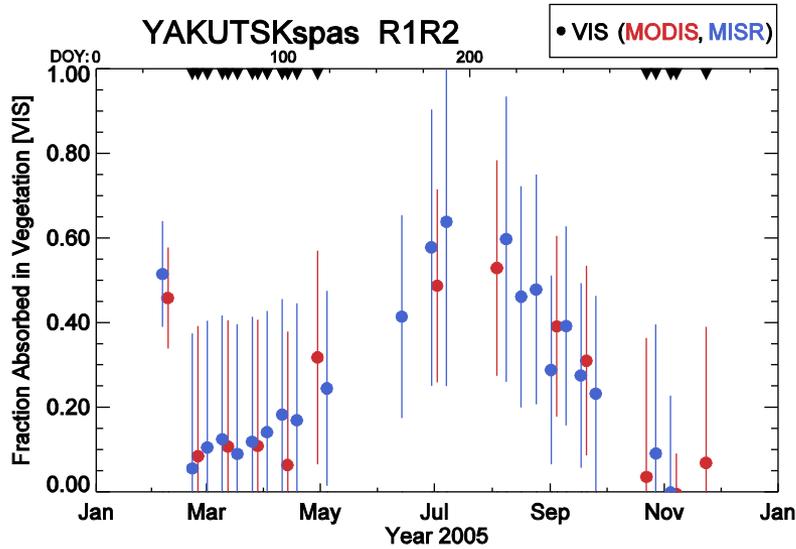
Application over Yakutsk: radiant fluxes



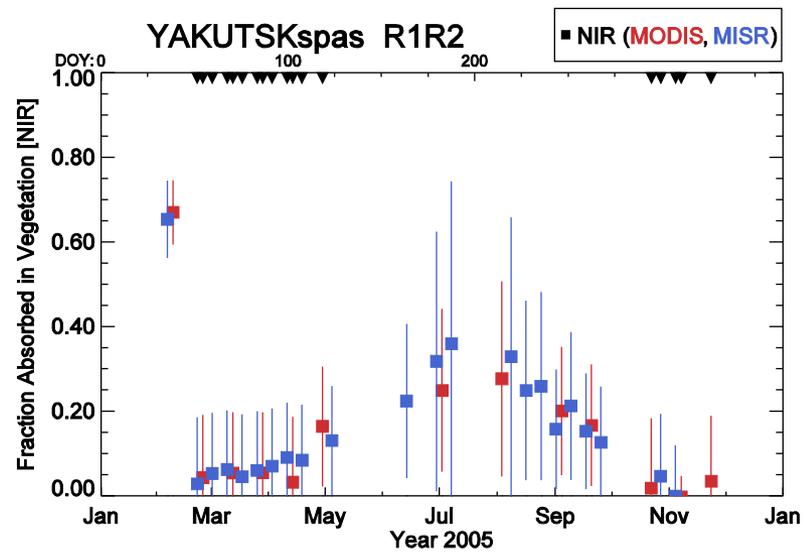
a priori 'green' leaves

Partitioning of the absorbed fluxes

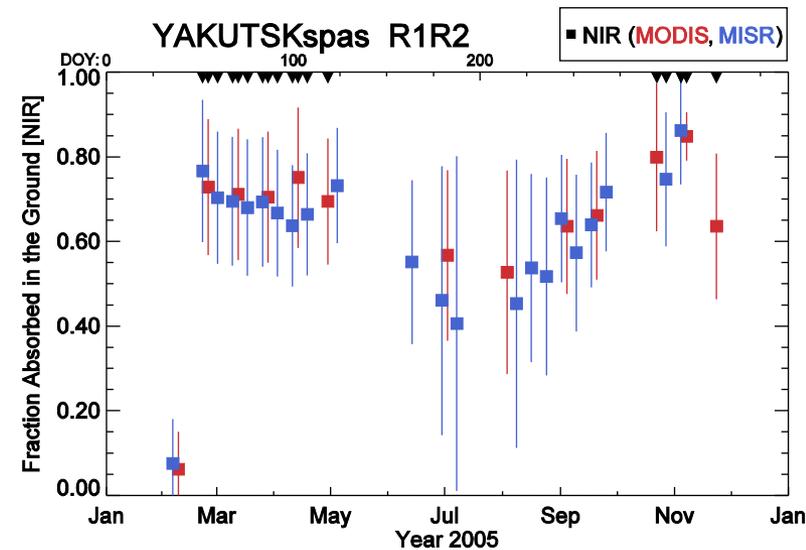
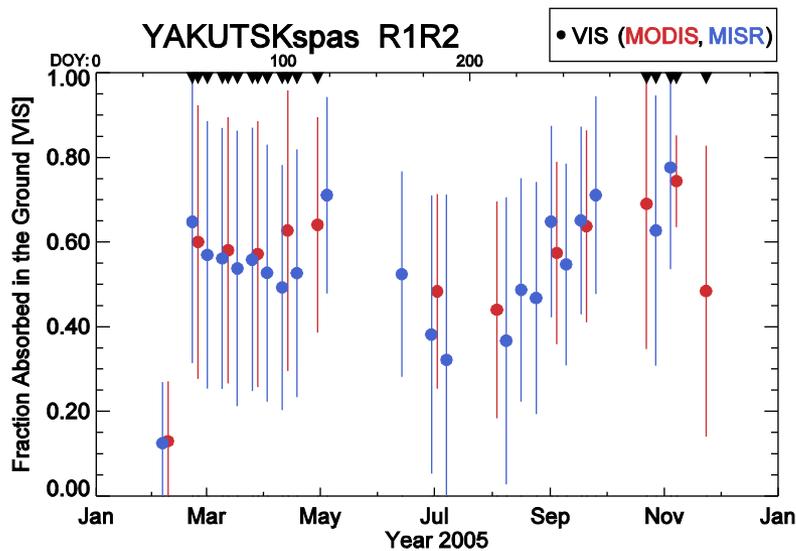
VISIBLE



NEAR-INFRARED



VEGETATION



SOIL

Add some examples over Northern region

Rukolahti period 161_16

map with $R=0.4$ $G=3.0$ $B=0.4$

case of agriculture:

61 56 06 and 28 55 29

Case of forest

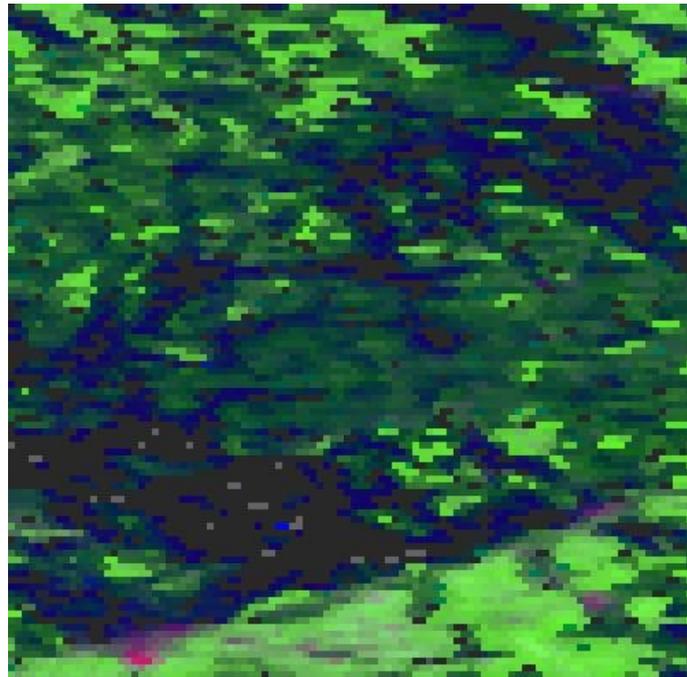
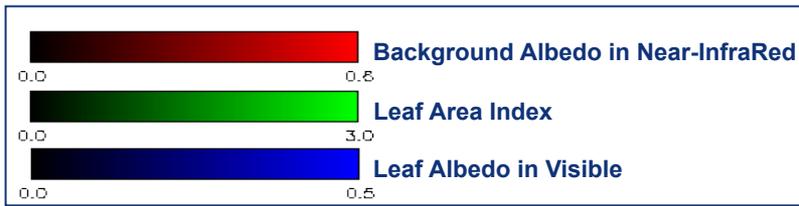
Then map period 241_16

With $R=0.4$ $G=0.8$ and $B=0.3$

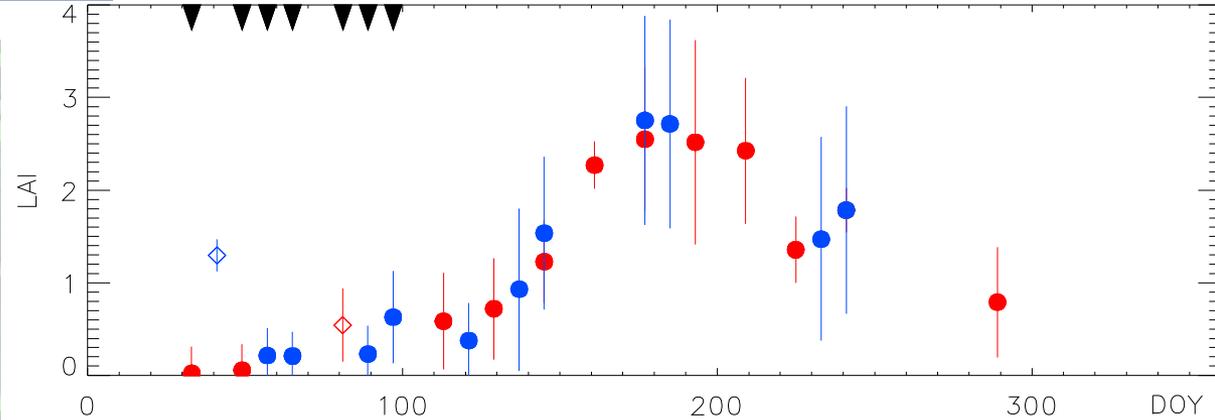
41 36 17 and 7 39 53

Application over Finland: Agriculture

Leaf Area Index

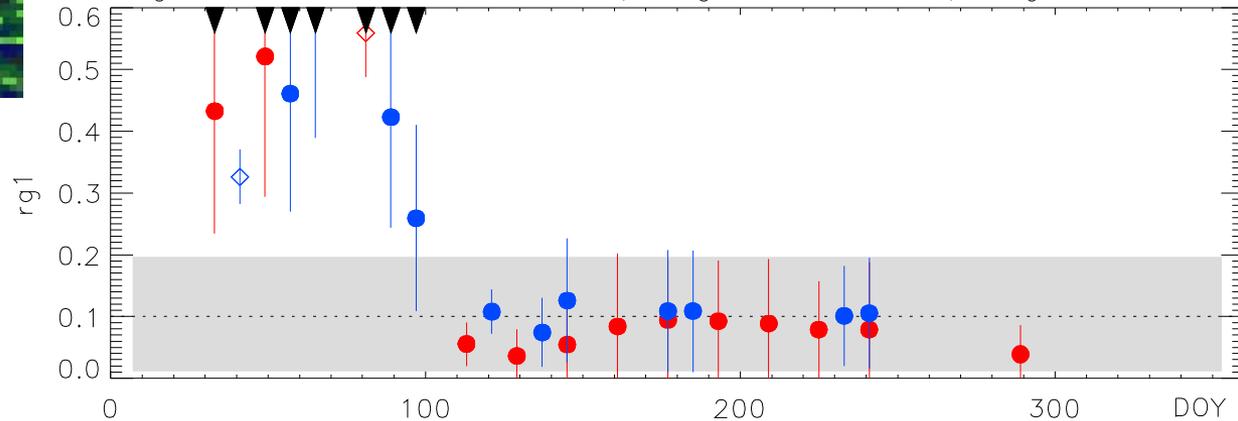


LAI RUOKOLAHTI: MODIS_RR/mrg2 - MISR_RR/mrg2 - 2005



Background albedo (visible)

rg1 RUOKOLAHTI: MODIS_RR/mrg2 - MISR_RR/mrg2 - 2005



Add some examples

Portugal period 145_16

map with $R=0.4$ $G=3.0$ $B=0.4$

case of agriculture:

41 53 42 and 7 49 29

Case of fire

Then map period 241_16

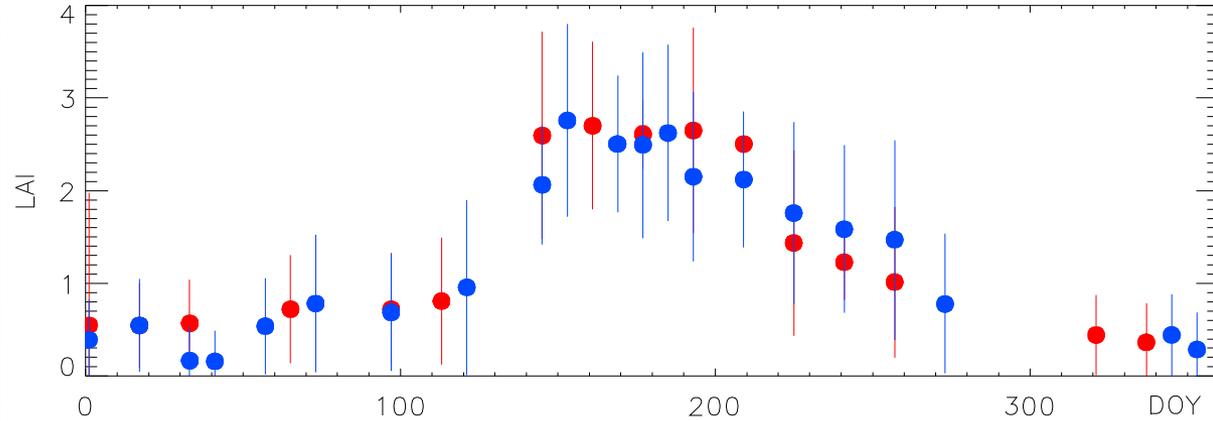
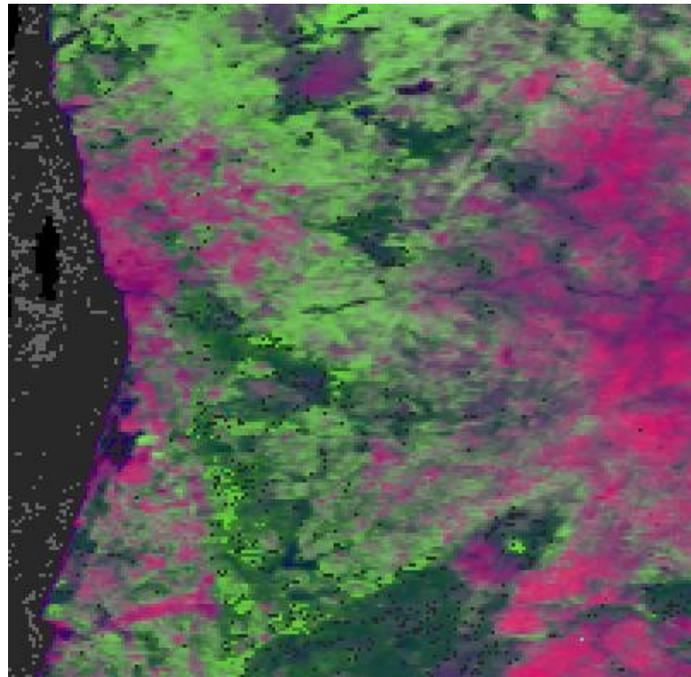
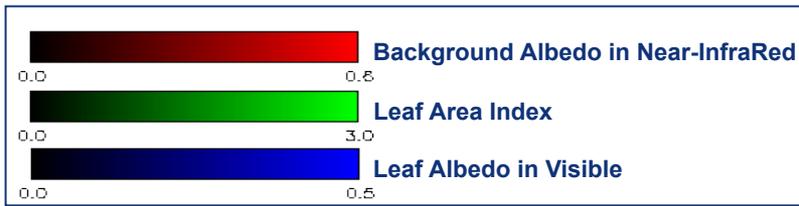
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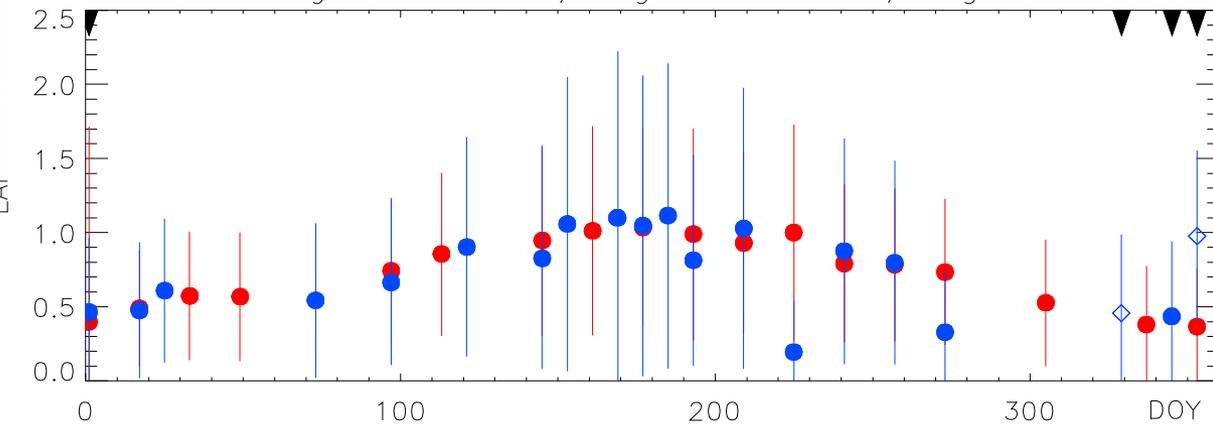
Application over Portugal: Agriculture

Leaf Area Index

LAI Portugal: MODIS_RR/mrg2 - MISR_RR/mrg2 - 2005



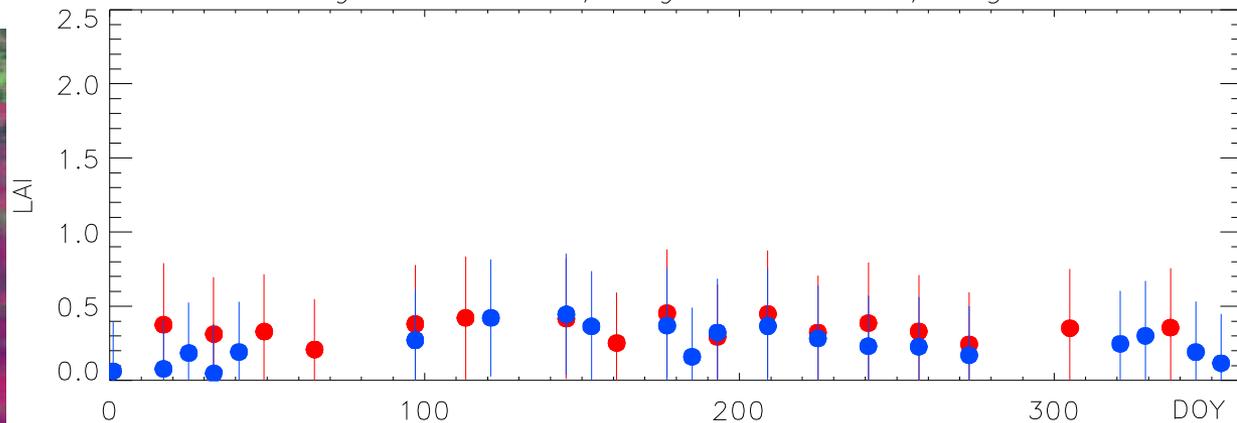
LAI Portugal: MODIS_RR/mrg2 - MISR_RR/mrg2 - 2005



Application over Portugal: sparse vegetation

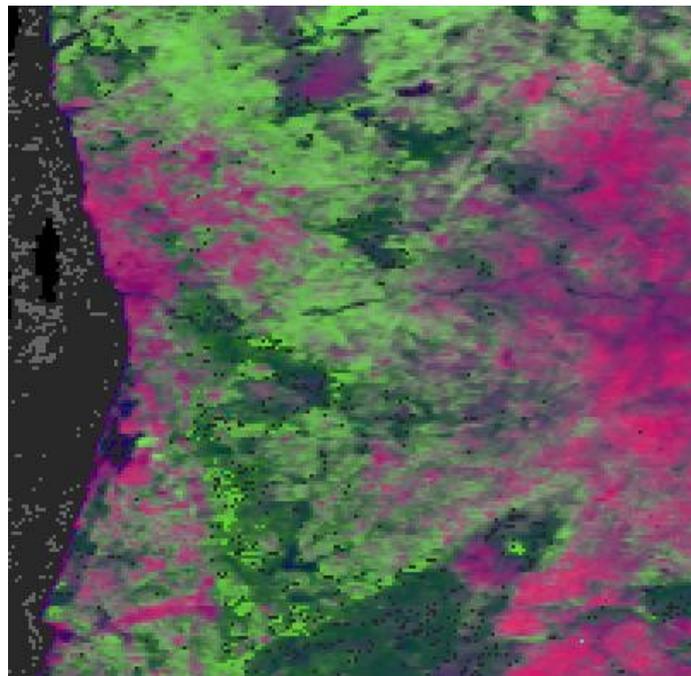
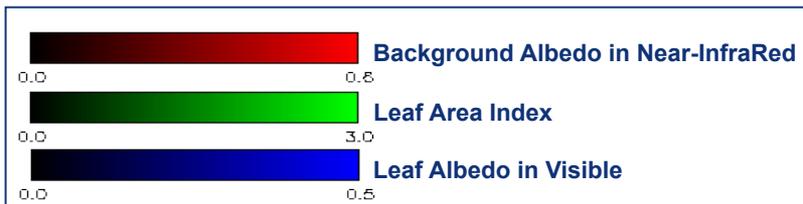
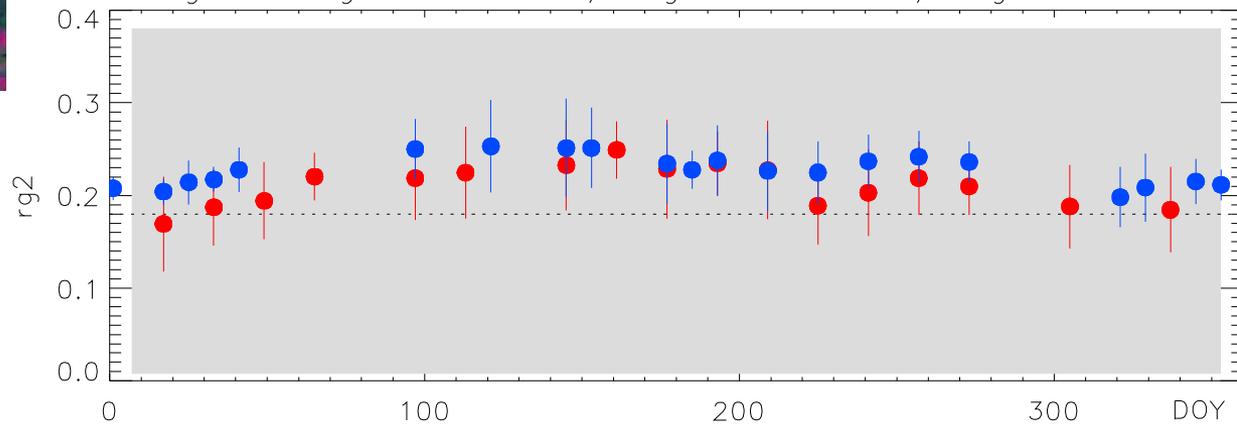
Leaf Area Index

LAI Portugal: MODIS_RR/mrg2 - MISR_RR/mrg2 - 2005



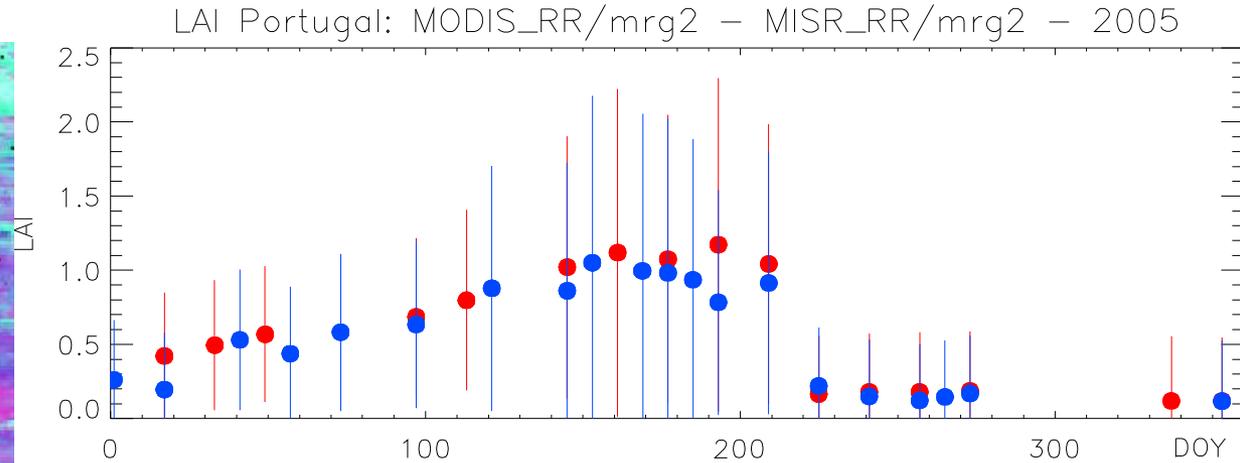
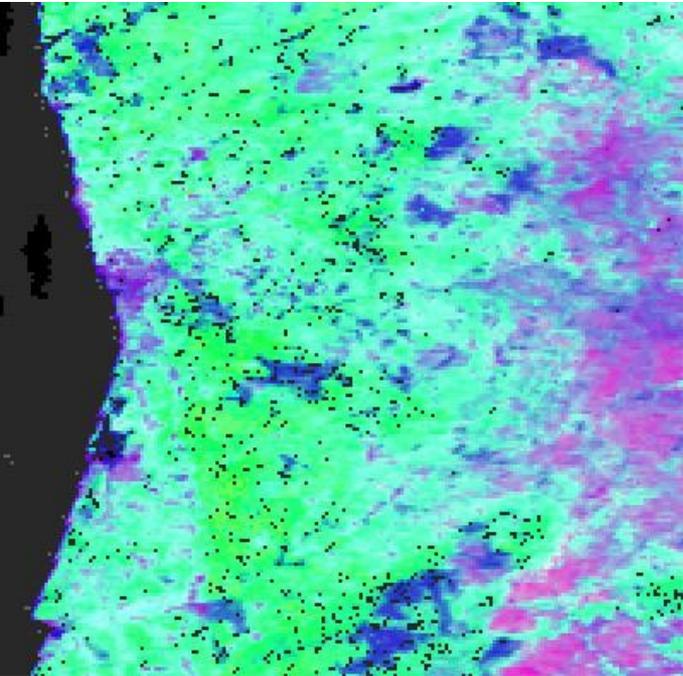
Background albedo (NIR)

rg2 Portugal: MODIS_RR/mrg2 - MISR_RR/mrg2 - 2005

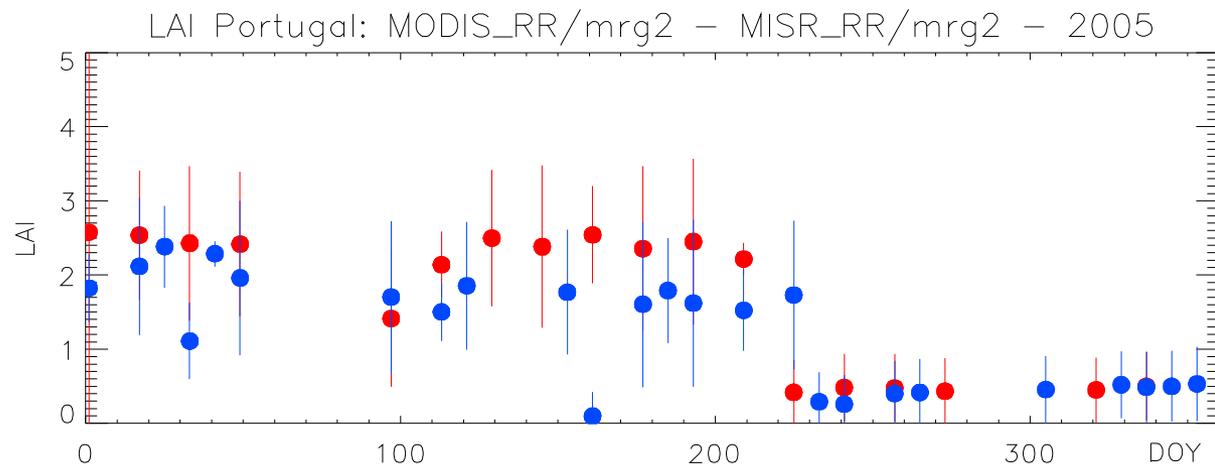


Application over Portugal: Burnt vegetation

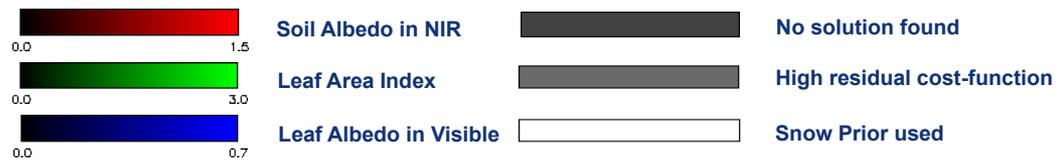
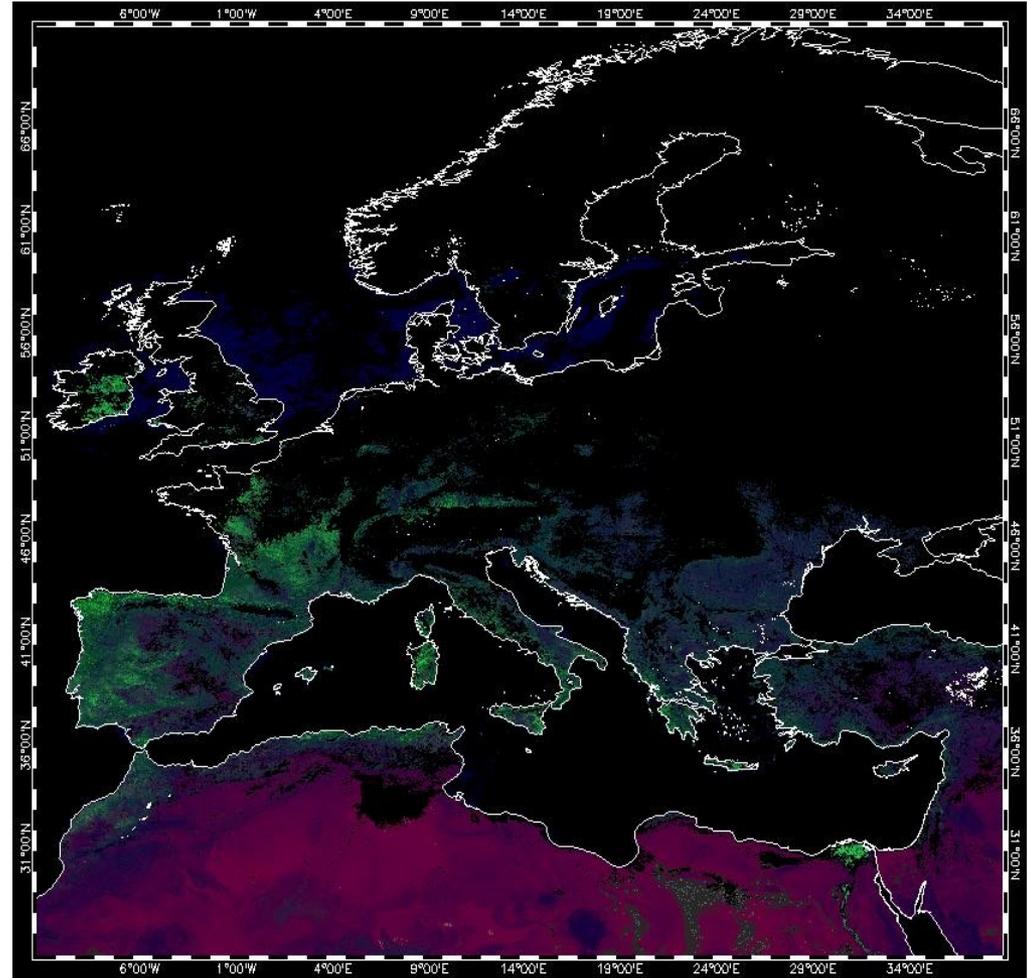
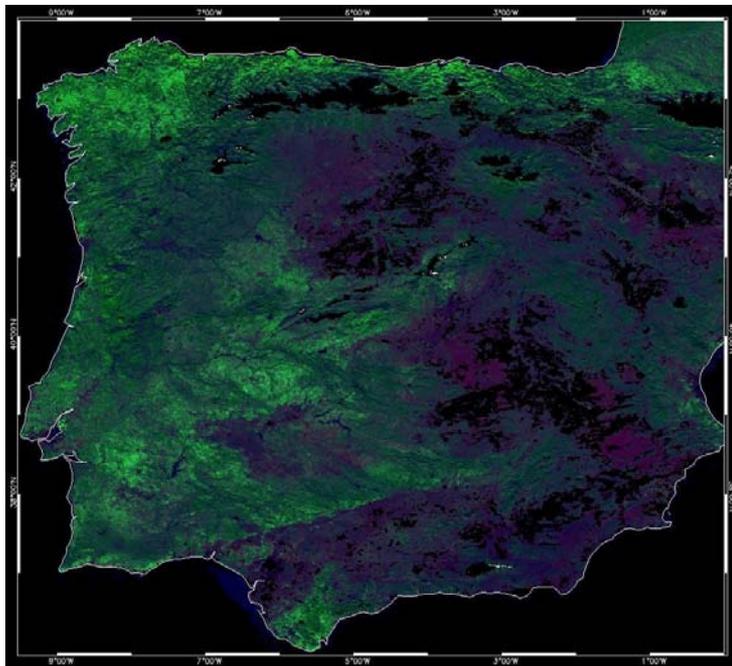
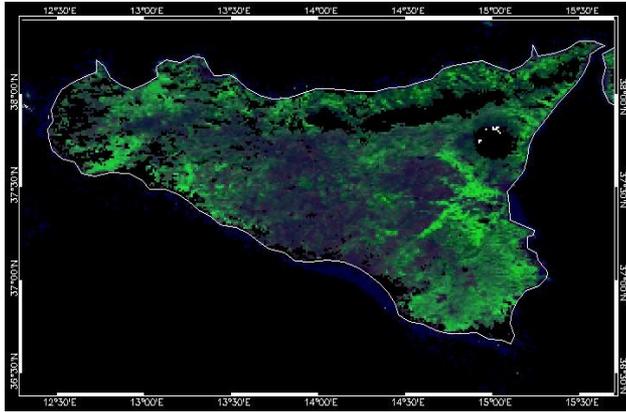
Leaf Area Index



Leaf Area Index



January 2005



Expected benefits

1. Derive **consistent set** of radiation fluxes in vegetation and soil layers (R, A and T + uncertainties) optimized against albedo products available from different sensors... and by-products.
2. Establish a quite **comprehensive baseline** against which historical records can be exploited.
3. Provide strong **predictive capabilities** for land surface monitoring needed for many applications.

Expected benefits

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 2. Establish a quite **comprehensive baseline** against which historical records can be exploited.
-
1. Derive land surface variables, e.g., LAI, leaf and soil colors to produce **Land Cover maps** under a controlled environment! ..new paradigm?
 1. Provide strong **predictive capabilities** for land surface monitoring needed for many applications.