



# Capitalizing on 1D RT models to interpret remote sensing data from 3D structurally heterogeneous vegetation systems

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# Context

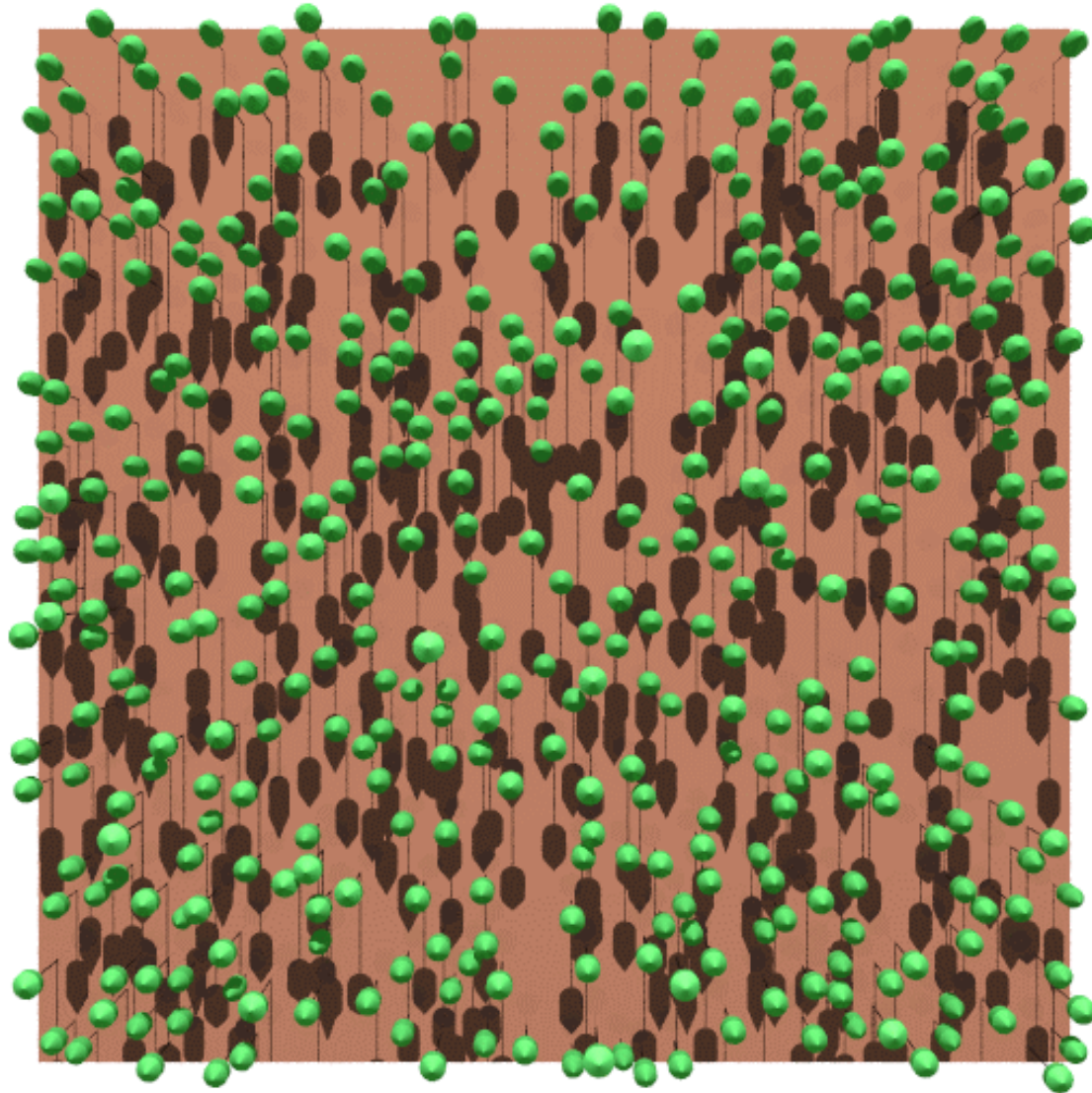
Interpret Remote Sensing data from **medium resolution optical sensors** to characterize (quantify) 3D heterogeneous vegetation (e.g. forests).

**MISR** (Terra) acquires multi-angular reflectances since 2000.

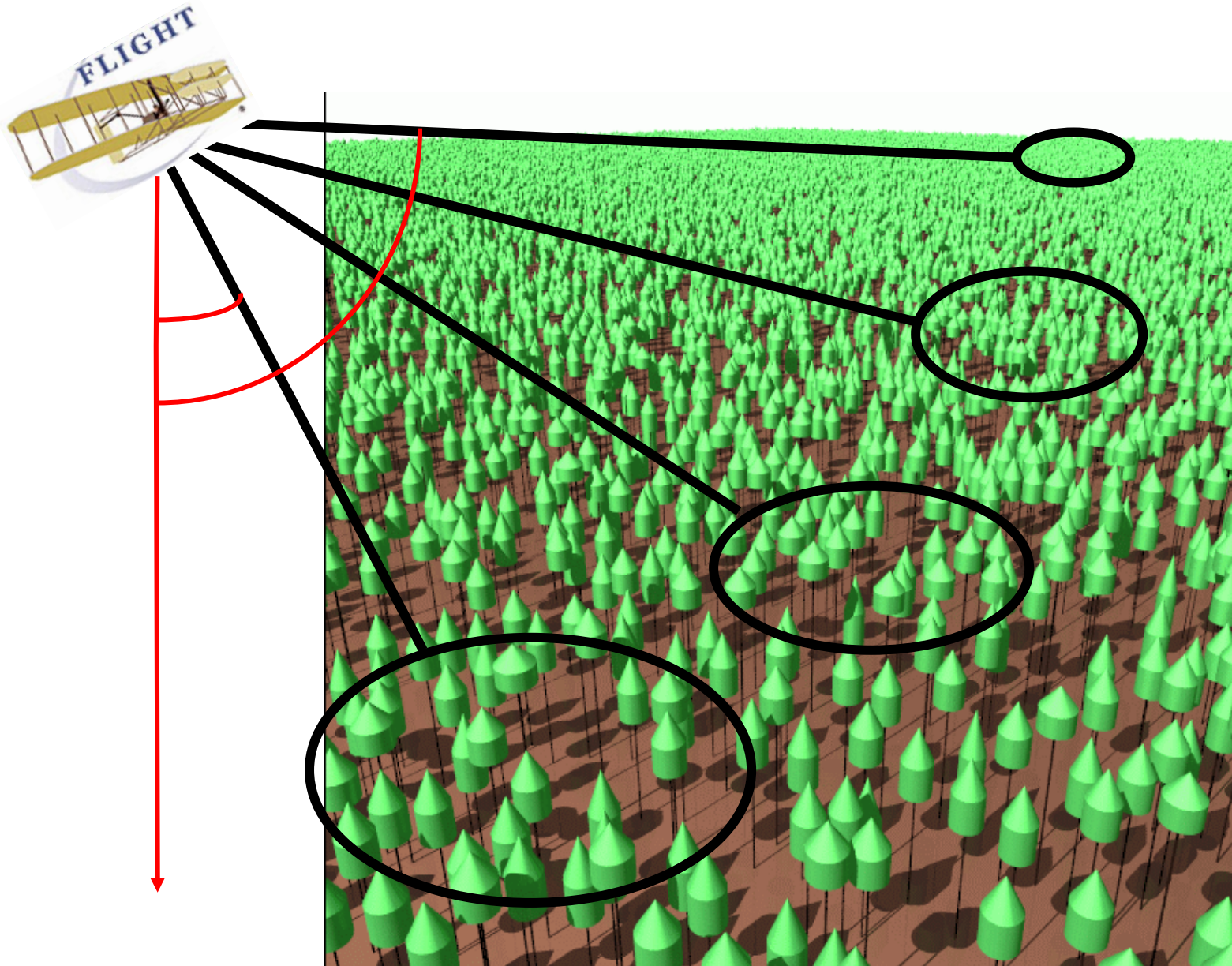
4 wavelengths, **9 view angles**

LandBRF (Level 2) are available at 1.1km resolution.

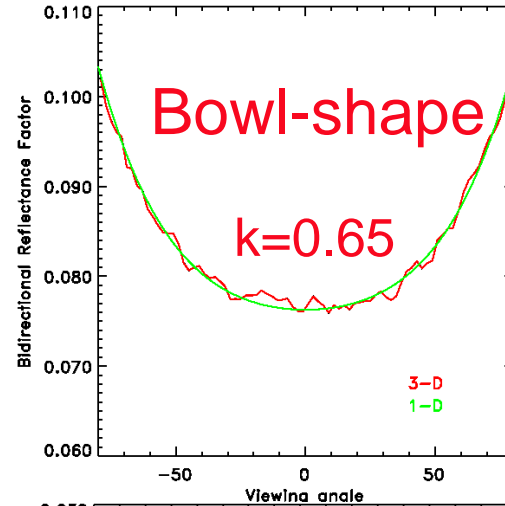
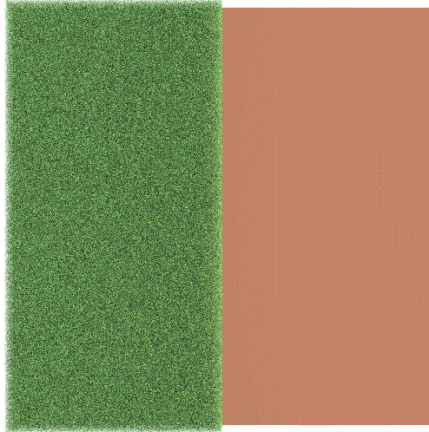
# Multiview views for surface structure



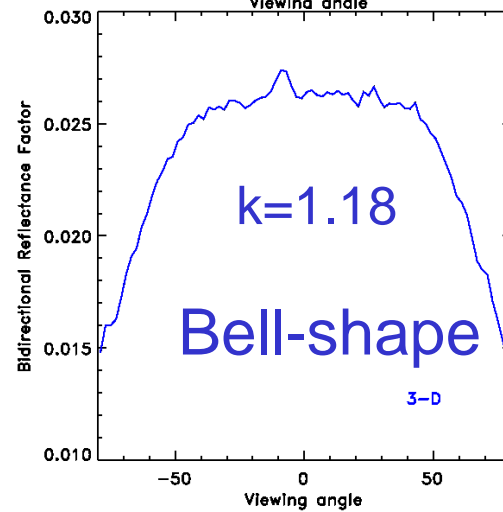
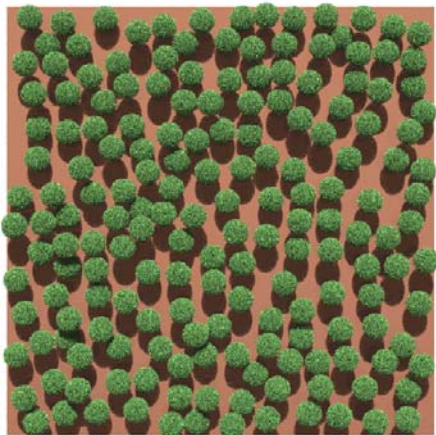
# Multiview views for surface structure



# Angular signatures in the red band



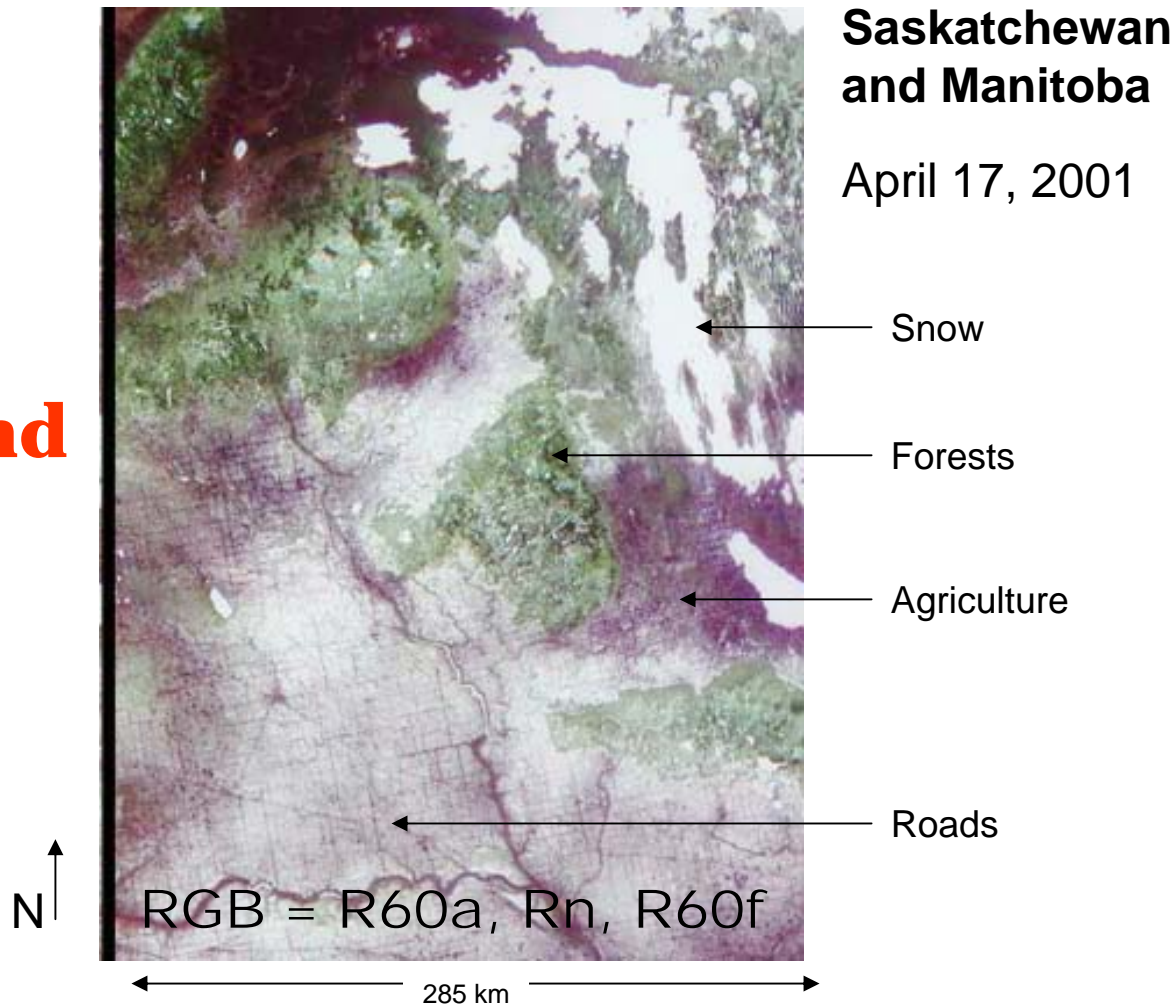
1-D



3-D

# Surface structure from MISR/Terra

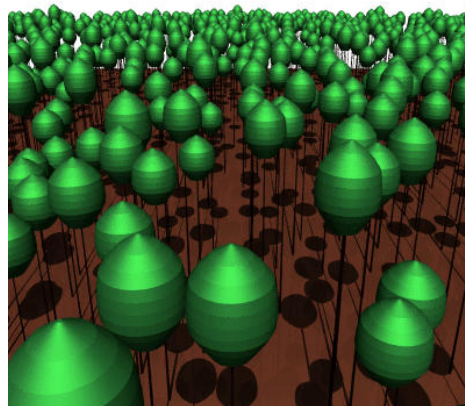
**Red band**



Ref: [http://www-misr.jpl.nasa.gov/gallery/galhistory/2001\\_may\\_30.html](http://www-misr.jpl.nasa.gov/gallery/galhistory/2001_may_30.html)

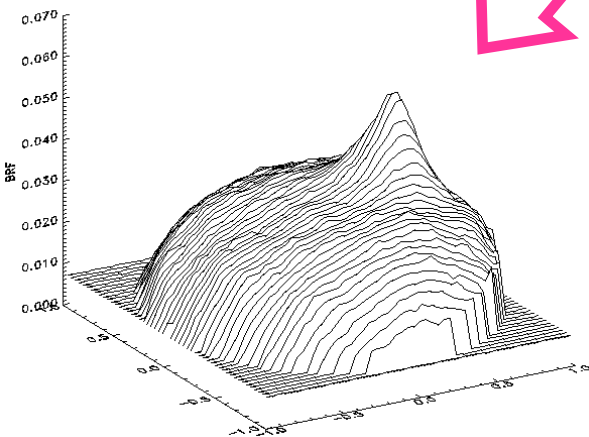
# Inversion problem

Biome parameters

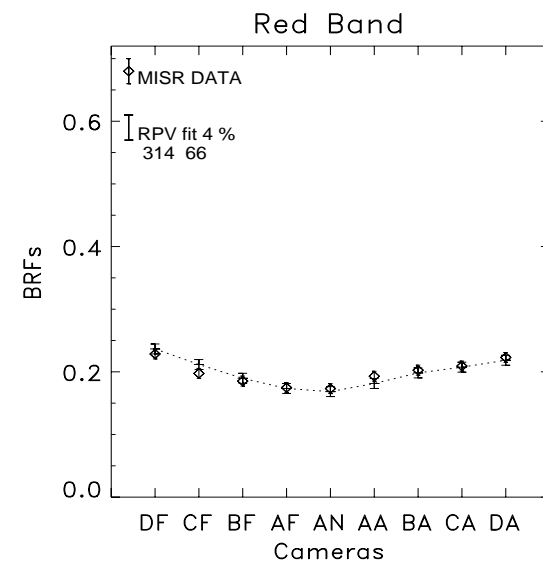


Retrieval

3-D radiative transfer model



Comparison



Simulated Bidirectional Reflectance Factor

Satellite BRF

# Modelling Issues

**3-D Monte-Carlo models are heavy to run.**

Pre-calculate **Look Up Tables (LUTs)** over many biomes:

**Forests:**

1) tree density	3) greenness of leaves
2) LAI in crowns	4) <i>etc</i>

+ background (anisotropic) reflectance

BRF to be simulated in (at least) 2 wavelengths,  
for several Sun angles and several view angles.



# Solutions

1) Have a fast(er) 3D MC model;

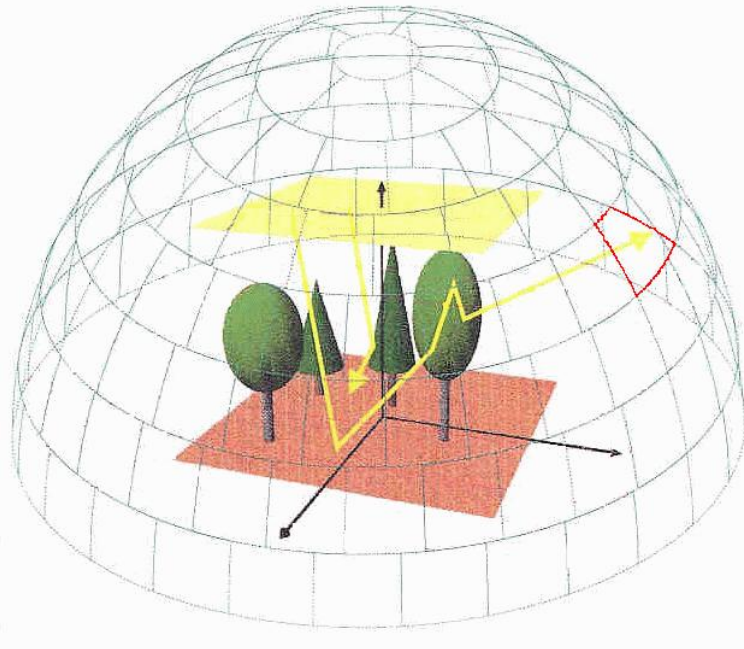
The Rayspread model

2) Save as much effort as possible in constituting the LUTs:

- a) decoupling background / vegetation;
- b) find spectrally invariant terms;
- c) simplify least significant contributions.

# The Rayspread model

Rayspread inherits from the Raytran model (3D Monte-Carlo raytracer)



Rayspread implements a variance-reduction technique: *local estimator* (aka *photon-spread*)

**~ 100 times faster.**

Rayspread results are non-discernible from Raytran (except for the Monte-Carlo noise). Both models participate to RAMI Phase 3.

# Solutions

1) Have a fast(er) 3D MC model;

## Radiative components

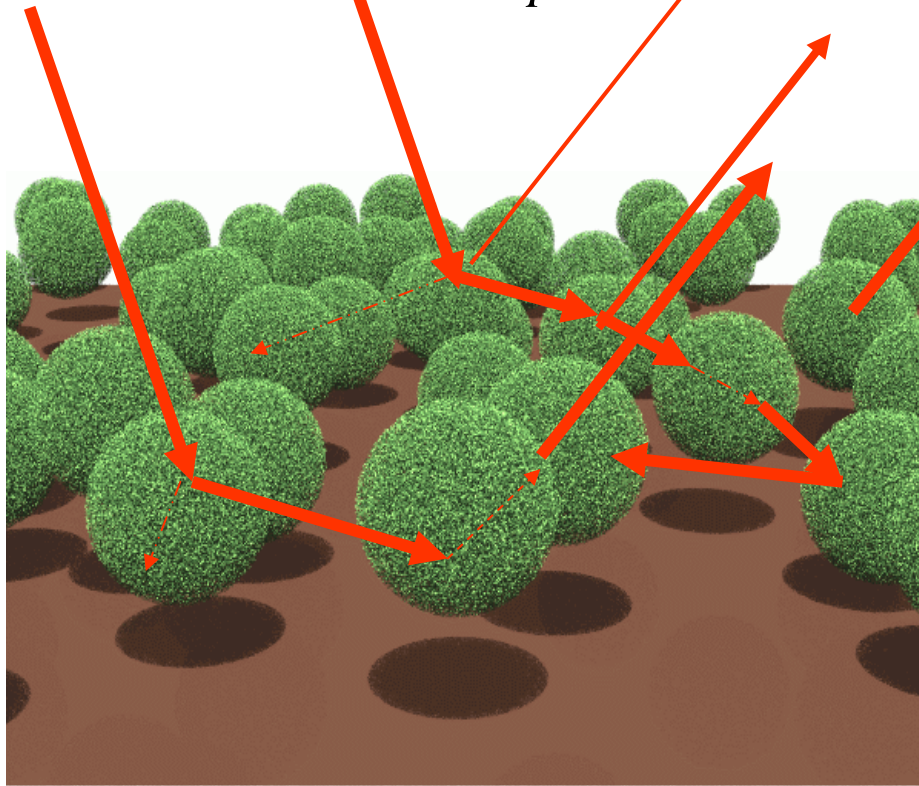
2) Save as much effort as possible in constituting the LUTs:

- a) decoupling background / vegetation;
- b) find spectrally invariant terms;
- c) simplify least significant contributions.

# Decompose the complex problem into simpler problems to solve



$$I_{coupled}^{\uparrow total} (Z_{toc}, \Omega_v, \Omega_0) = I_{vegetation}^{\uparrow Collided} (Z_{toc}, \Omega_v, \Omega_0)$$



**Black Background contribution**

- 1) It regulates the *absorption* processes associated to vegetation photosynthesis
- 2) Strongly depending on the *density* of green vegetation

# Decompose the complex problem into simpler problems to solve



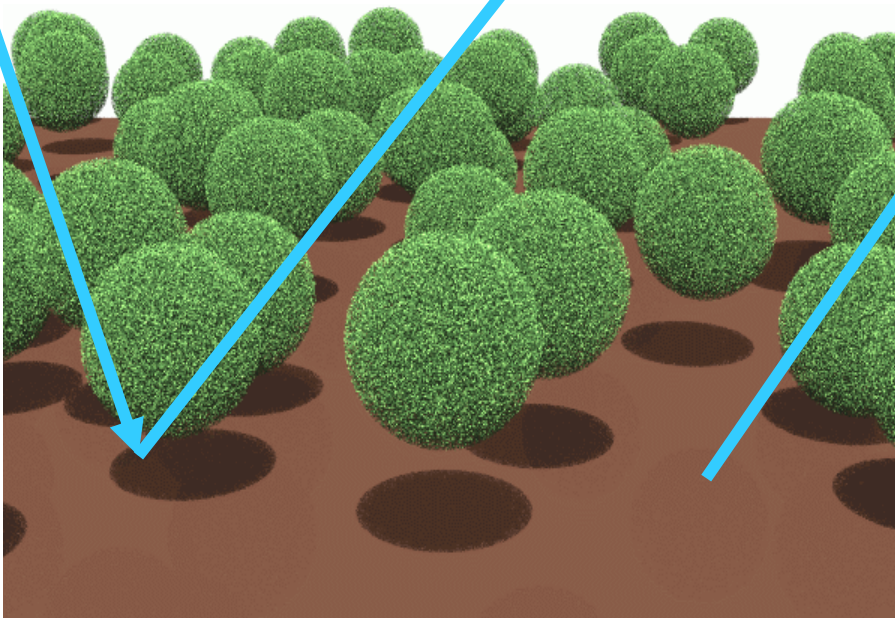
$$I_{coupled}^{\uparrow total}(Z_{toc}, \Omega_v, \Omega_0) = I_{vegetation}^{\uparrow Collided}(Z_{toc}, \Omega_v, \Omega_0)$$

$$+ I_{background}^{\uparrow Uncollided}(Z_{toc}, \Omega_v, \Omega_0)$$

**Black Canopy contribution**

1) No absorption process by vegetation associated with this **wavelength independent contribution**

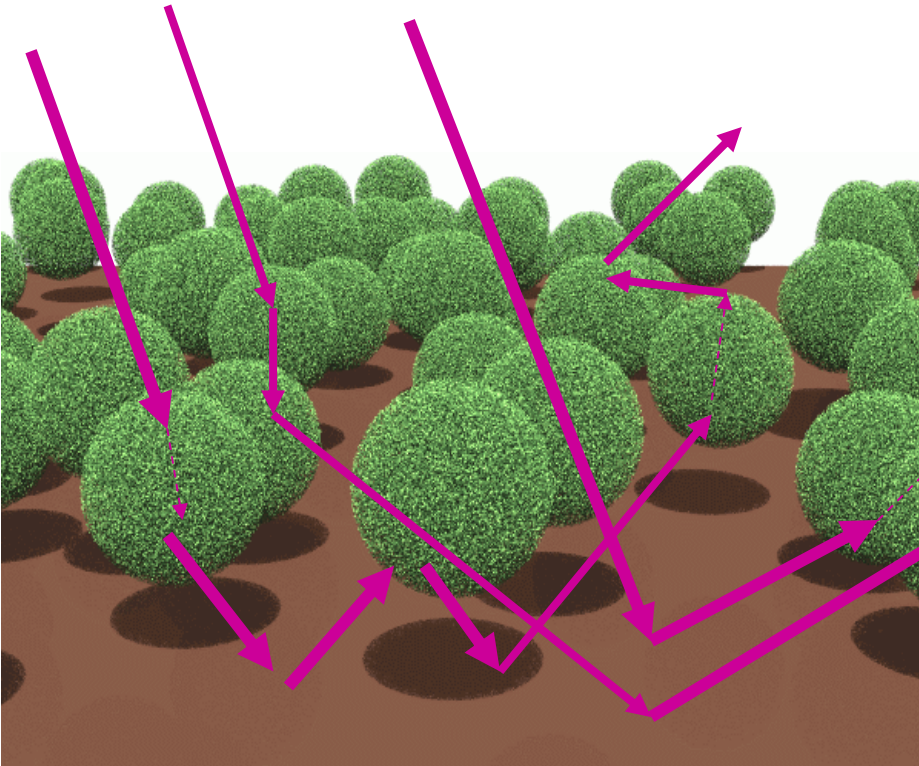
2) Strongly controlled by **3-D distribution of vegetation extinction coefficient**



# Decompose the complex problem into simpler problems to solve



$$I_{coupled}^{\uparrow total}(Z_{toc}, \Omega_v, \Omega_0) = I_{vegetation}^{\uparrow Collided}(Z_{toc}, \Omega_v, \Omega_0)$$



$$+ I_{background}^{\uparrow Uncollided}(Z_{toc}, \Omega_v, \Omega_0)$$

$$+ I_{background}^{\uparrow Multiplecollided}(Z_{toc}, \Omega_v, \Omega_0)$$

1) Controlled by multiple scattering events between the background and the canopy

2) Mostly **negligible** in the visible domain of the solar spectrum

# Use a 1-D solution for the background-canopy multiple scattering contribution

$$3^{\text{rd}} \text{ term} = \int_{\text{background}}^{\uparrow \text{MultipleCollided}} (Z_{\text{toc}}, \Omega_{\mathbf{v}}, \Omega_0)$$

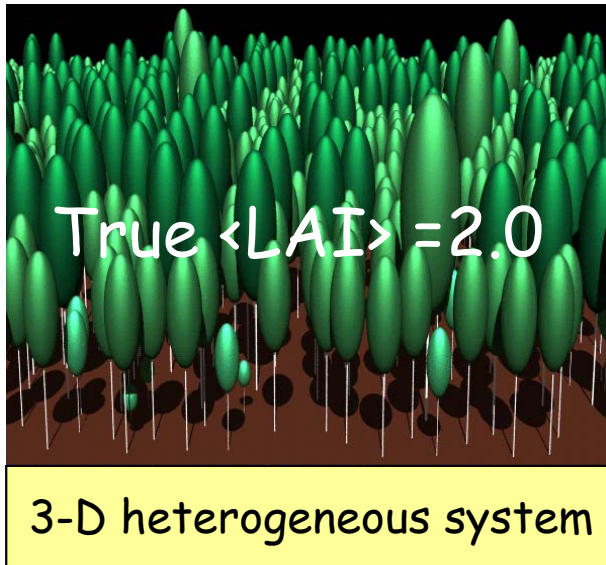
Developed analytical formulation of this intensity.

Assume a homogeneous plane-parallel, turbid medium representation (1-D solution).

"Effective" state variable values are required for

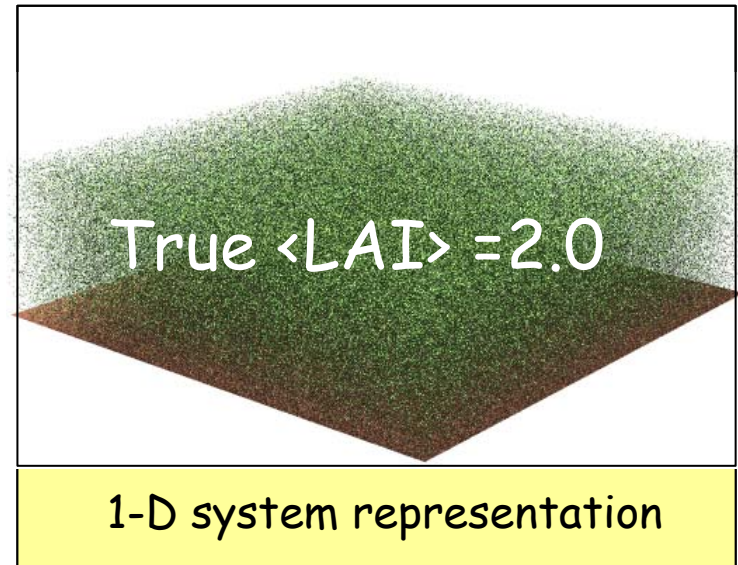
the 1-D model:  $\widetilde{LAI}$ ,  $\widetilde{r}_l$  and  $\widetilde{t}_l$

# Effective state variable values ensure that 1-D fluxes are close to 3-D fluxes.



Direct transmission at 30 degrees Sun zenith angle,

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

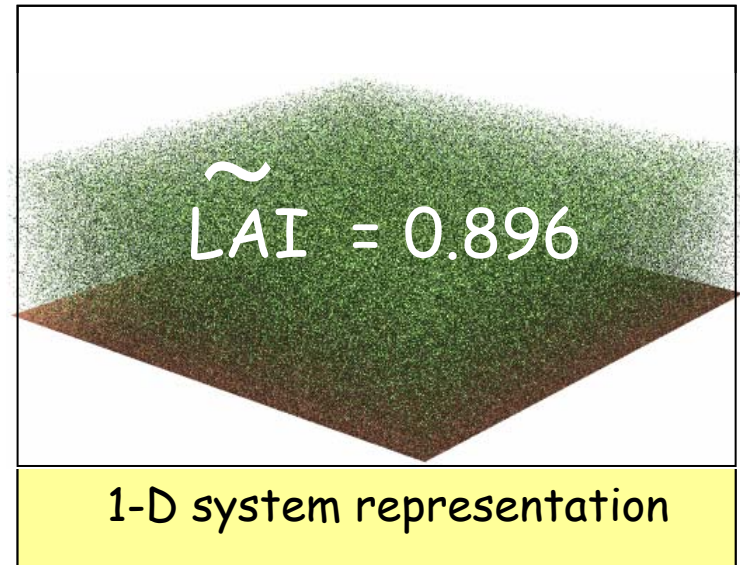
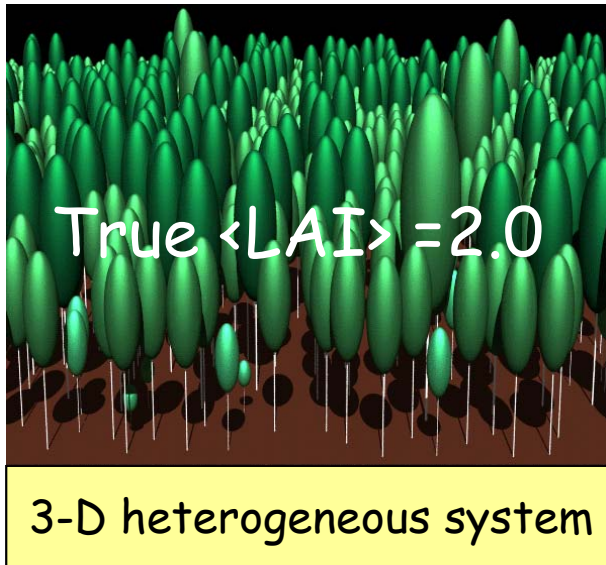


Direct transmission at 30 degrees Sun zenith angle,

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



Effective state variable values ensure that  
1-D fluxes are close to with 3-D fluxes.



$$\tilde{LAI}(\mu_0) = \langle LAI \rangle \cdot \zeta(\mu_0)$$

# Estimates of the effective quantities

$$\tilde{r}_l \text{ and } \tilde{t}_l$$

are retrieved together from inversion of the **R** and **Tdiff** fluxes associated with the **Black Background** contribution from 3-D MC simulations.

$$\sum_{\mu_0} |\overline{R}_{veg}^{Coll}(z_{toc}, \mu_0; \mathbf{B}, \mathbf{O}, \mathbf{S}, LAI_{scene})| - \overline{R}_{veg}^{||Coll}(z_{toc}, \mu_0; \tilde{r}_l, \tilde{t}_l, \widetilde{LAI}(\mu_0)|$$

Reflected flux

$$\sum_{\mu_0} |\overline{T}_{bgd1}^{Coll}(z_{bgd}, \mu_0; \mathbf{B}, \mathbf{O}, \mathbf{S}, LAI_{scene})| - \overline{T}_{bgd1}^{||Coll}(z_{bgd}, \mu_0; \tilde{r}_l, \tilde{t}_l, \widetilde{LAI}(\mu_0)|$$

Transmitted flux

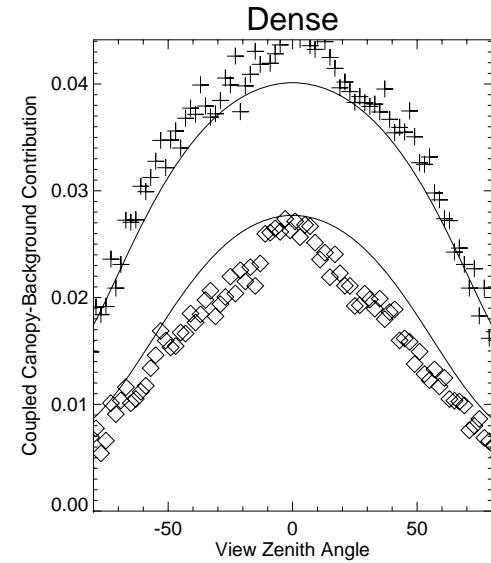
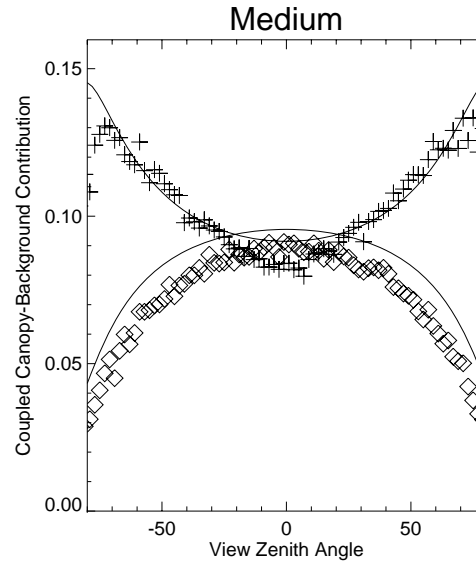
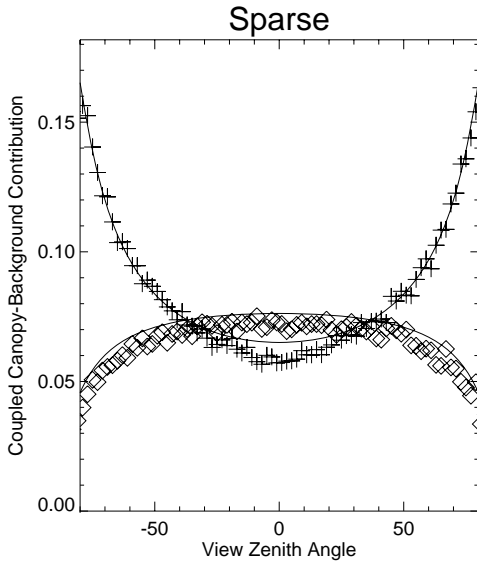
# Ratio of effective (1-D)/true (3-D) values

Variable	Sparse	Medium	Dense
$\tilde{LAI}$ (30°)	0.35	0.45	0.76
$\tilde{r}_l + \tilde{t}_l$ (NIR)	0.82	0.84	0.88
$\tilde{r}_l / \tilde{t}_l$ (NIR)	4.65	5.76	7.14

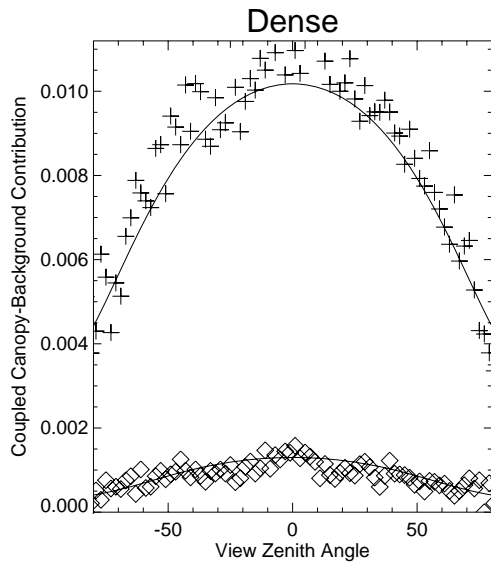
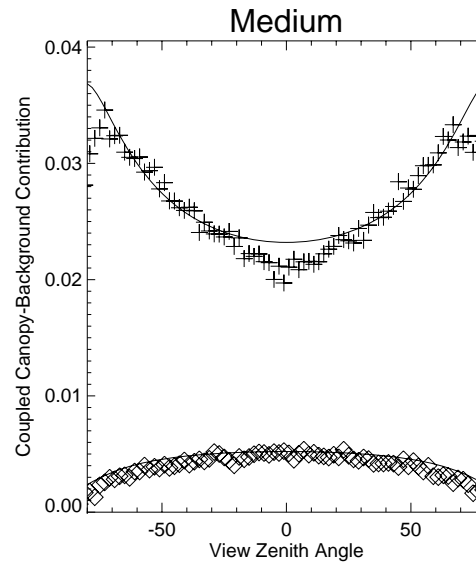
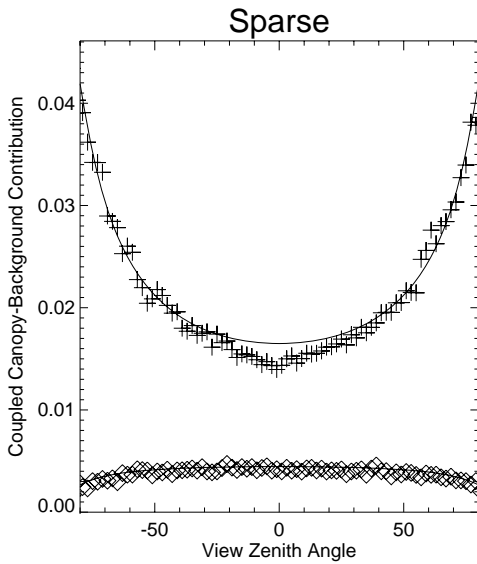
Effective variables are retrieved so that the 1-D model reproduces (at best) the 3-D MC **fluxes**.

Use the 1-D model for the estimation of directional intensities: **BRFs**

# Performance of the analytical solution



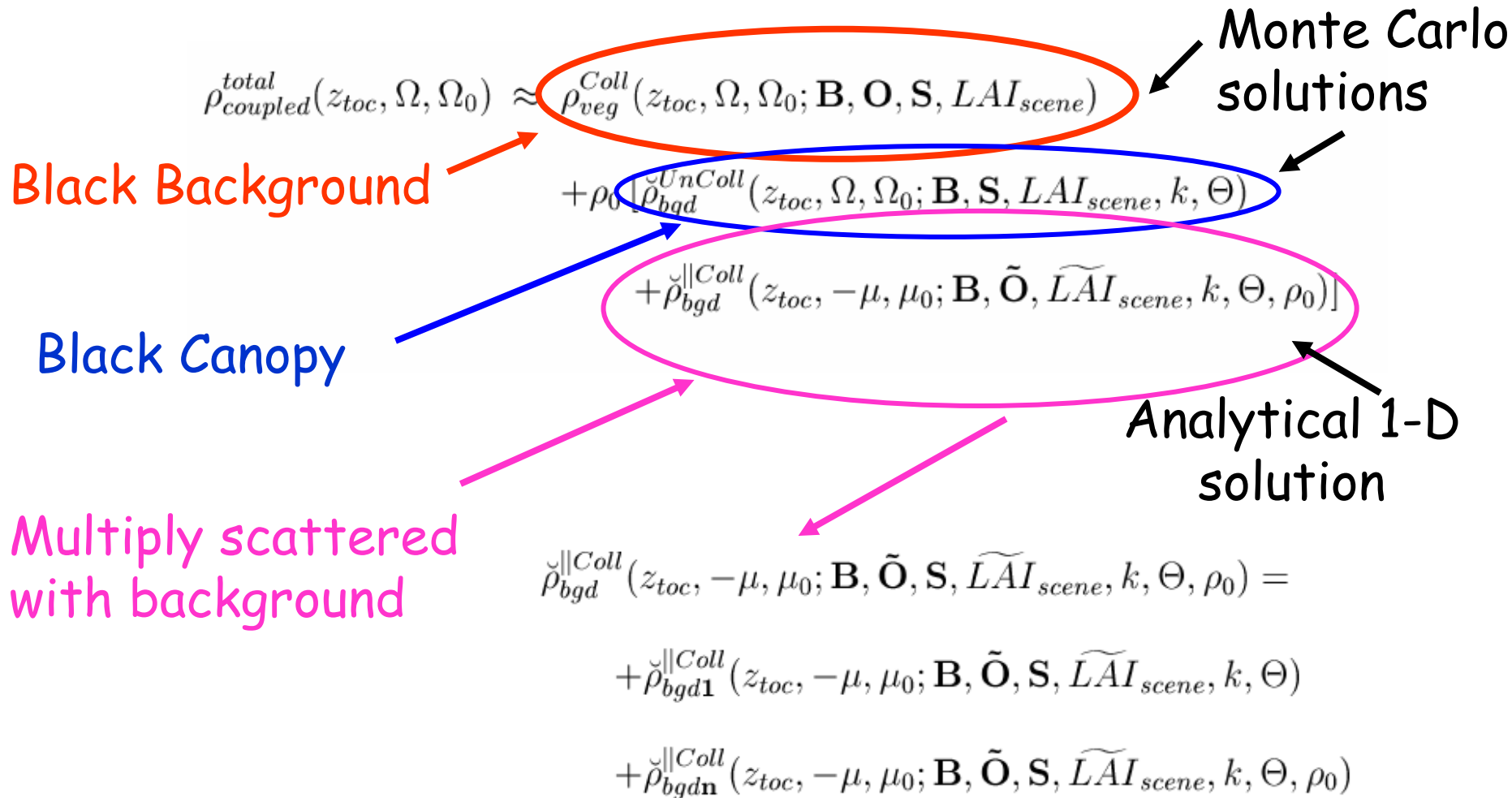
Snow-covered background



Typical background

Sun angle at  $30^\circ$

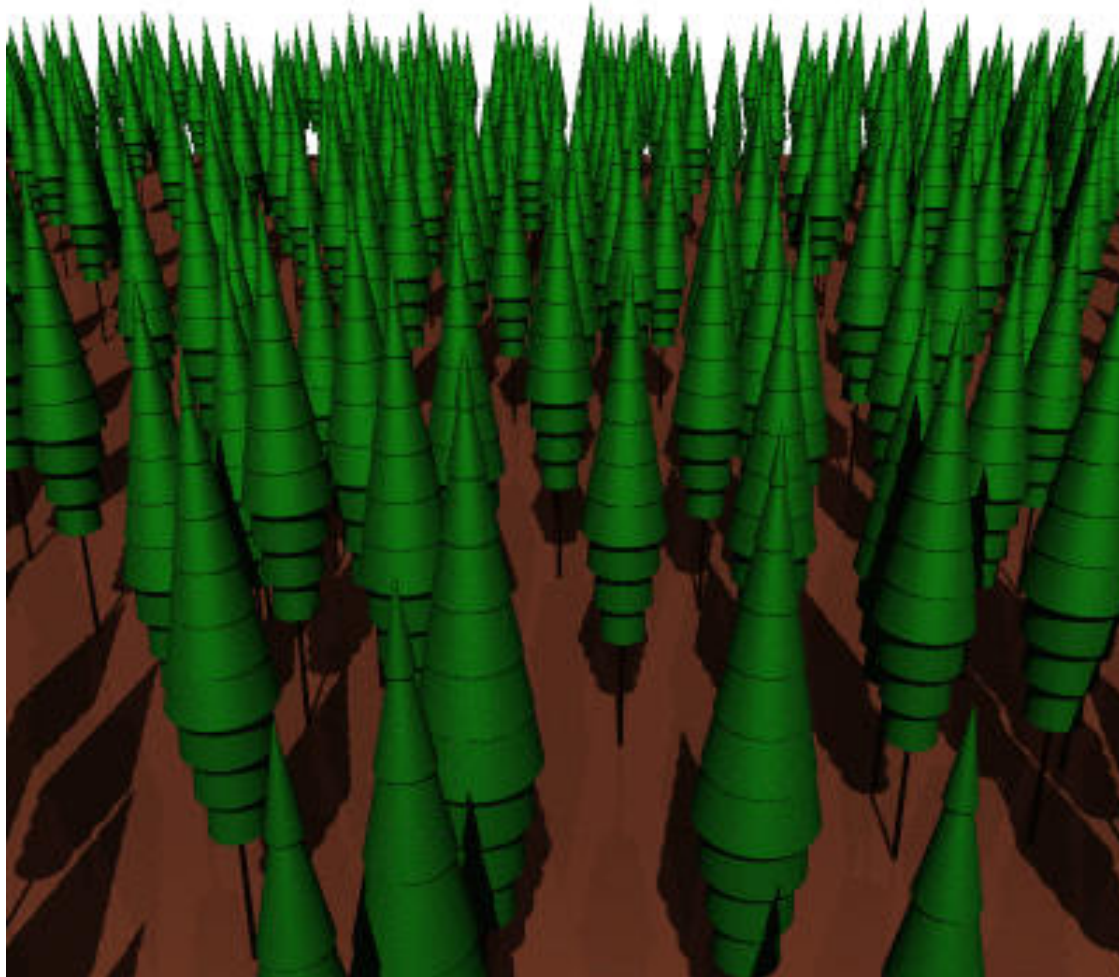
# Final simplified expression for the radiance fields from any complex land surface



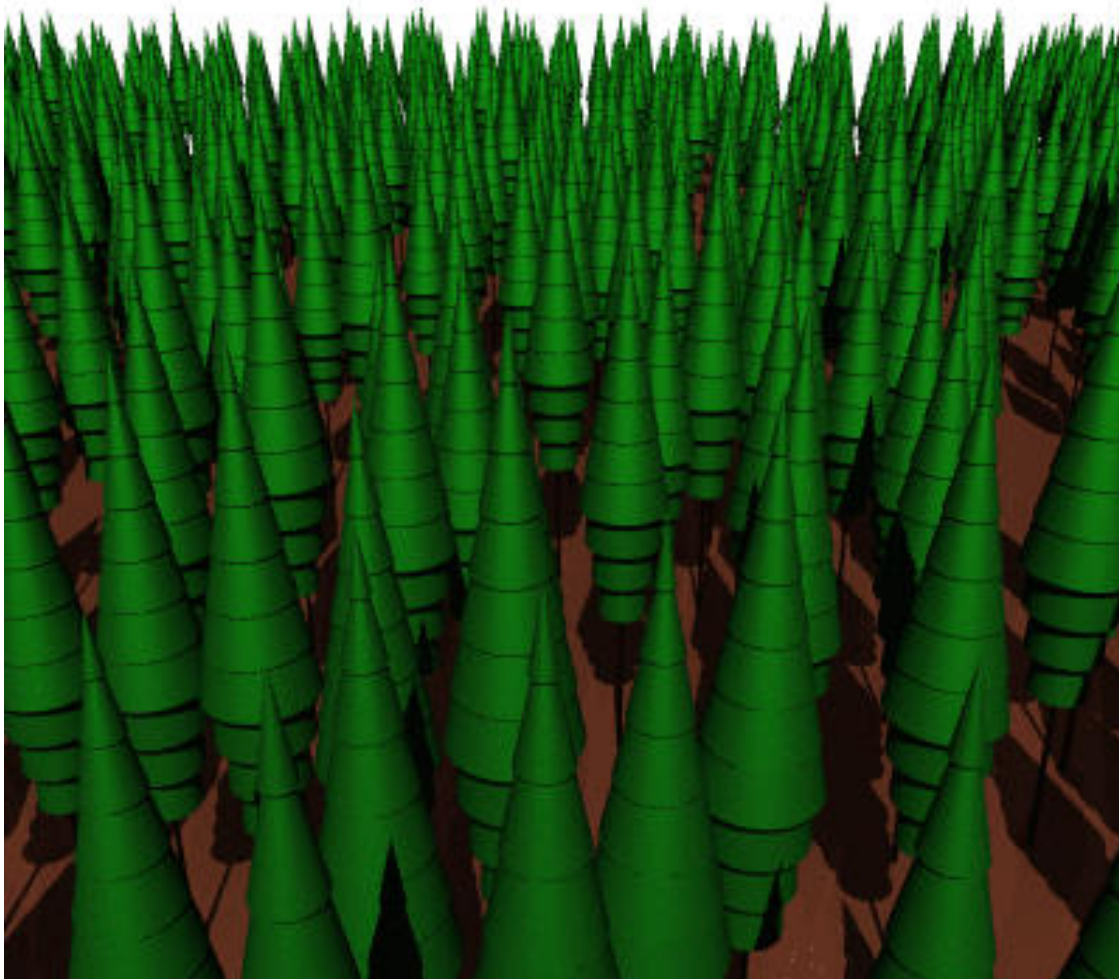
# Inversion strategy

- Use ecological knowledge, e.g., allometric equations to design a large set of canopy scenarios once for all (ref. Widlowski et al. (2003))

# Examples of geophysical scenarios

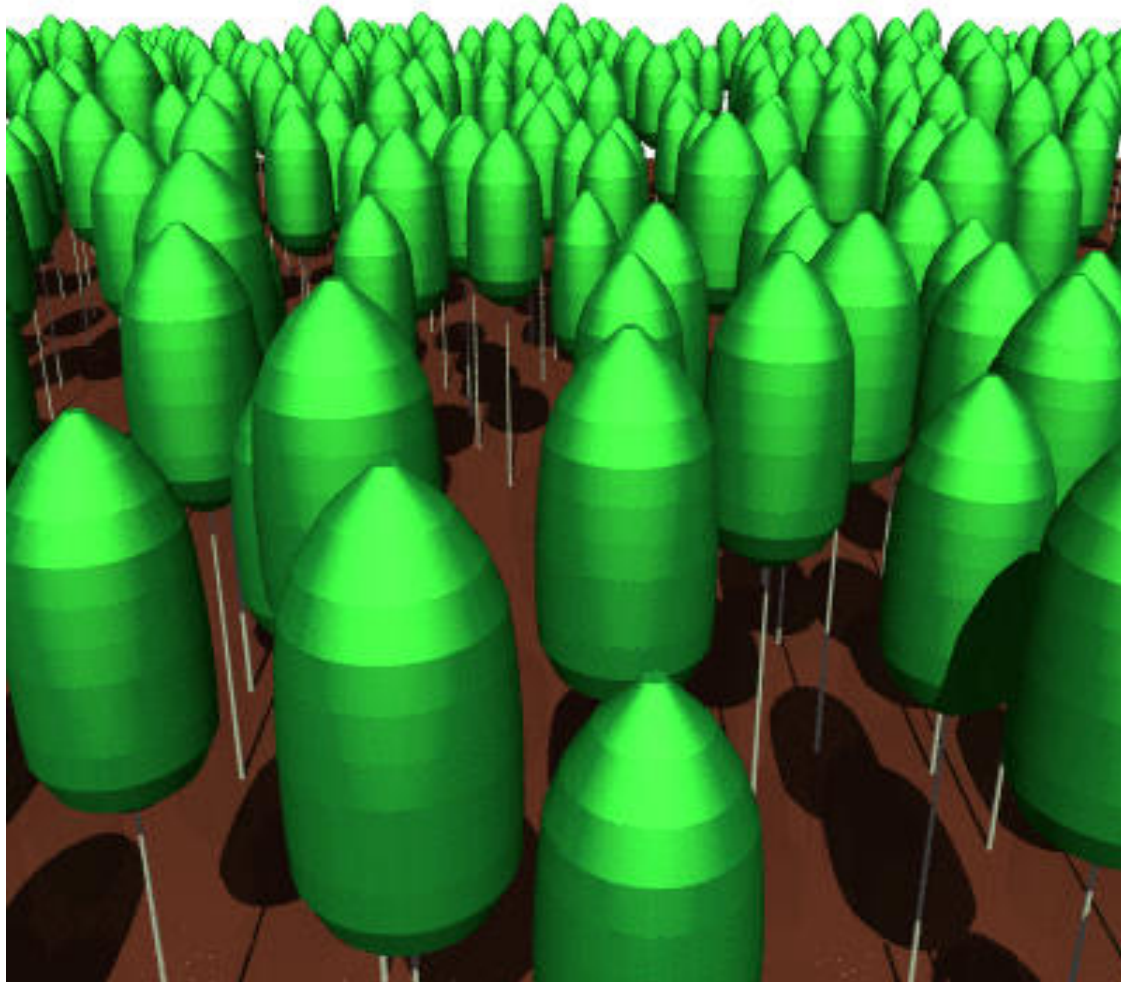


# Examples of geophysical scenarios

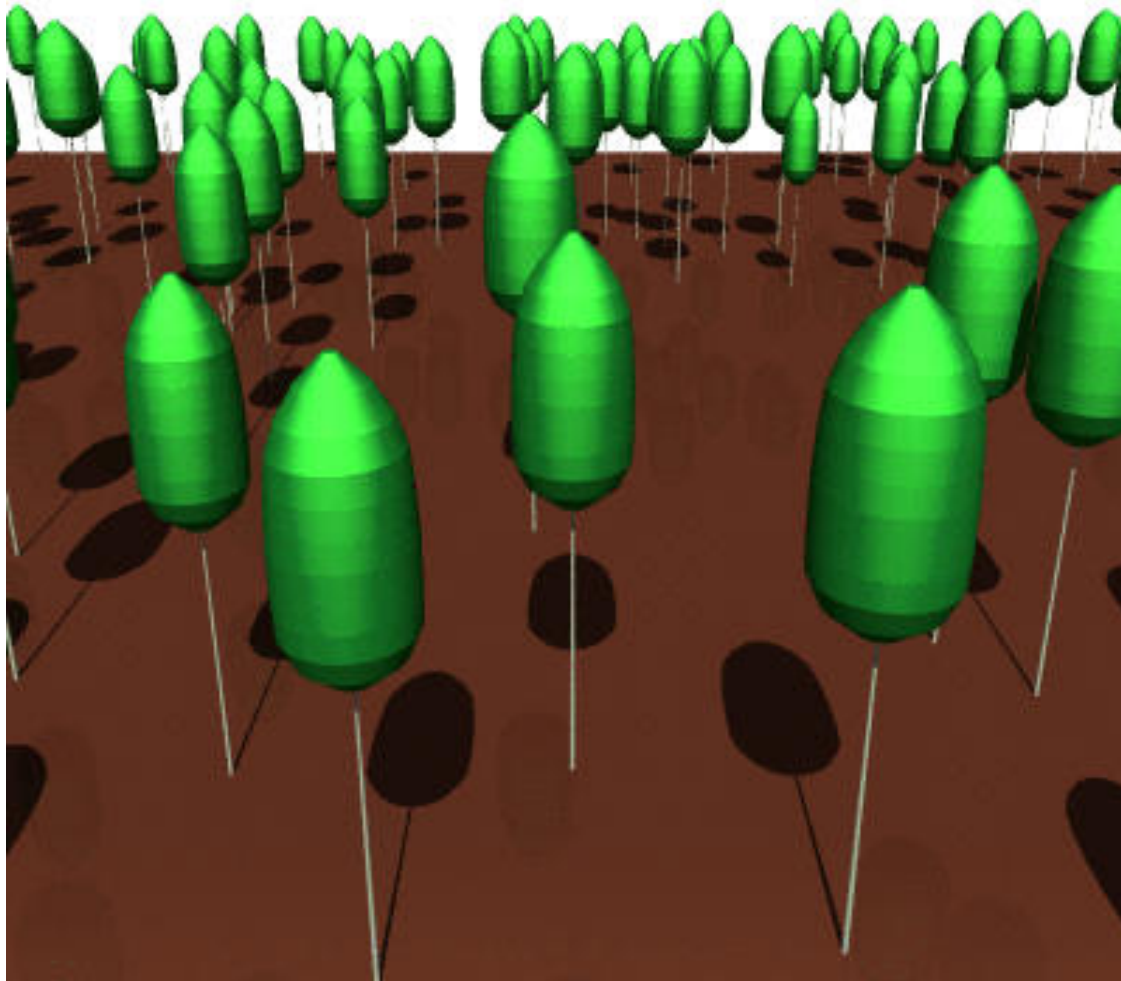




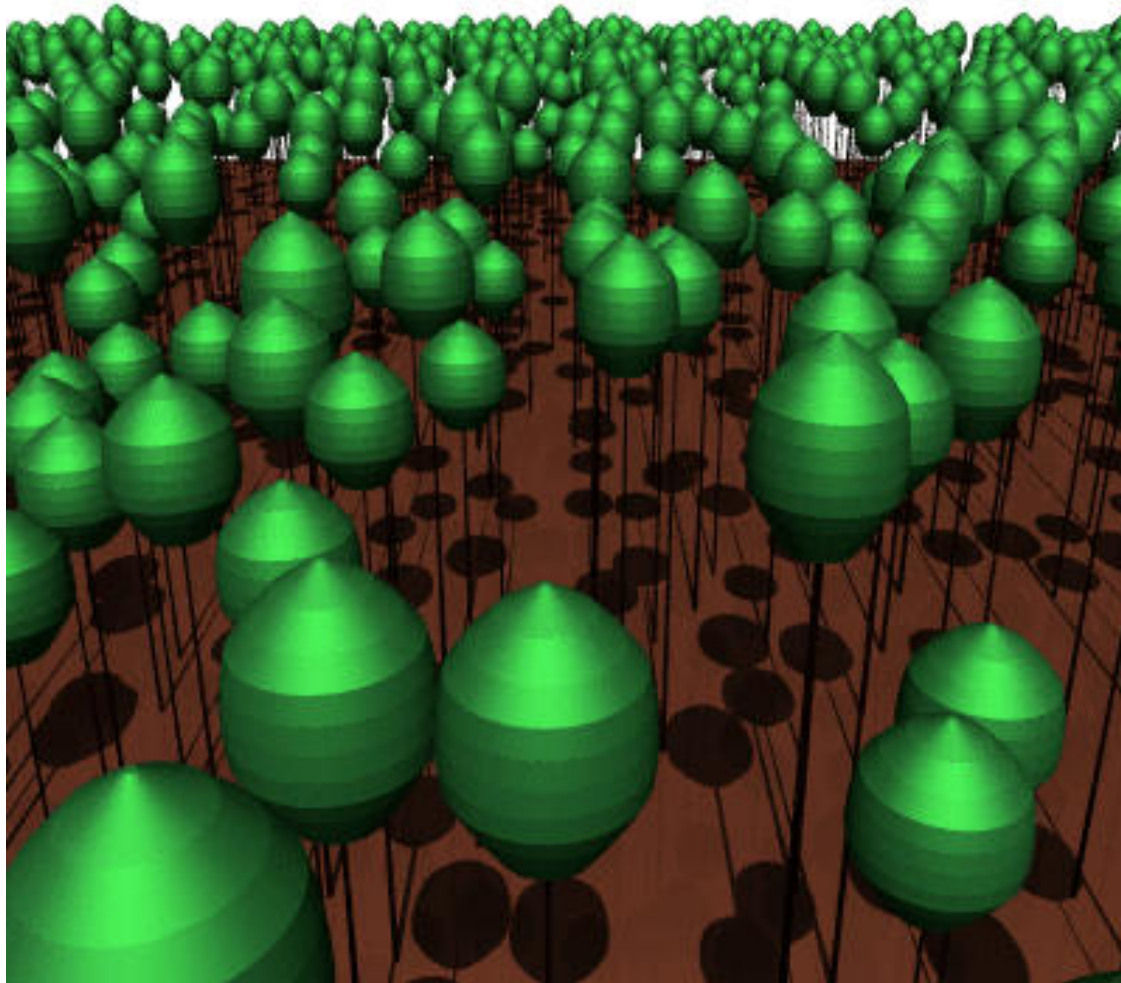
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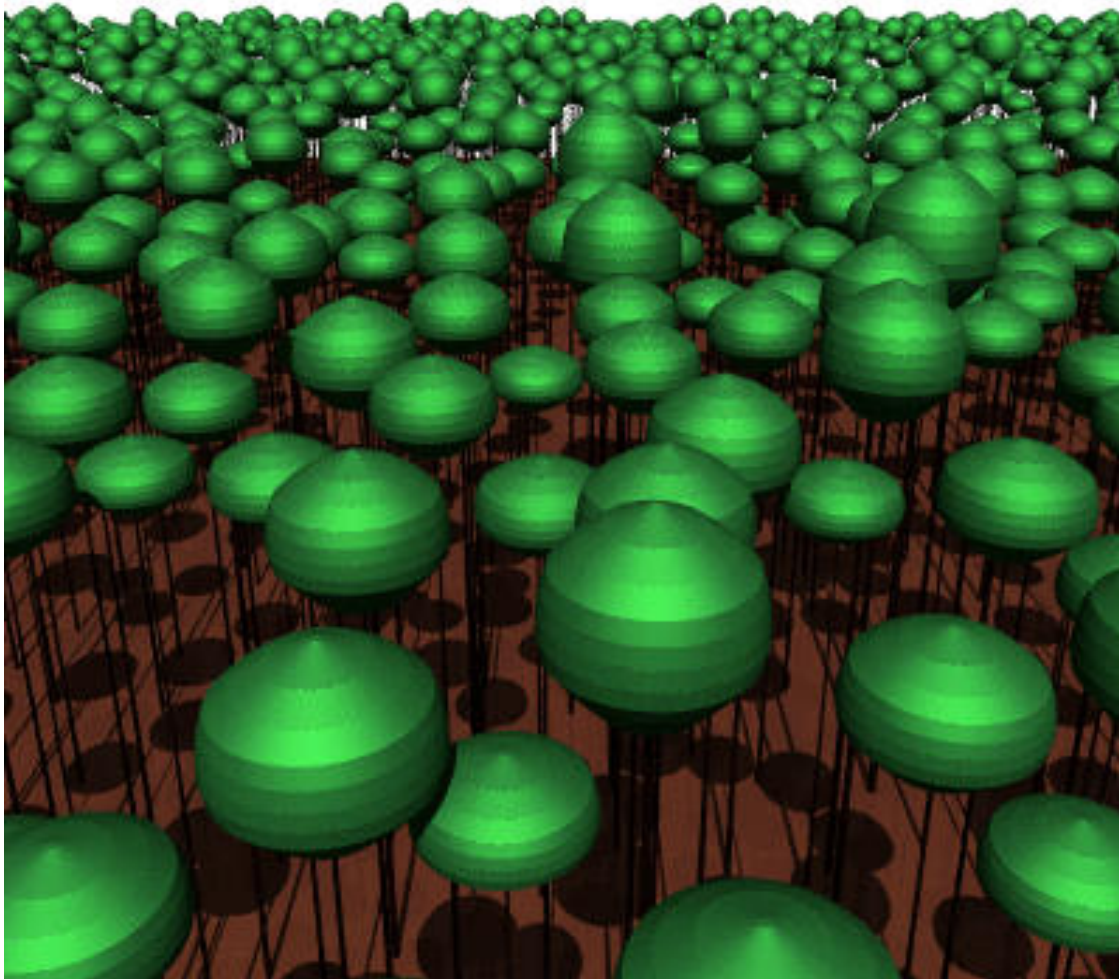
# Examples of geophysical scenarios



# Examples of geophysical scenarios



# Examples of geophysical scenarios



# Inversion strategy

- Use ecological knowledge, e.g., allometric equations to design a large set of canopy scenarios once for all (ref. Widlowski et al. (2003))
- Estimate the **Black Background** and **Black Canopy** contributions associated with all scenarios: use 3-D MC simulations.
- Estimate the effective state variable values of the 1-D problem in order to simulate efficiently the **vegetation-background coupled contribution** (1-D model is faster).
- Adopt an inversion scheme minimizing the distance between measurements and simulations.

# What did we gain?

- Inversion scheme based on limited MC simulations: **Black Background** and **Black Canopy** components only are requested.
- The **background brightness** value is solved as a continuous variable during the inversion.
- Allows detection of fast changes in the background conditions (e.g., snow and snow melting) and slow modifications in the canopy (e.g. LAI,...)



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Thank you... Questions?

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<http://www-gem.jrc.it/stars/>