



DART: a 3D radiative transfer model for studying natural and urban surfaces

Yingjie Wang^a, O. Regaieg^{a, b}, N. Lauret^a, J. Guilleux^a, E. Chavanon^a, Z. Zhen^{a, c}, P. Boitard^a, R. Démoulin^a, T. Nguyen^{a, b}, H. Jin^d, X. Yang^d, T. Yin^e, A. Mkaouar^f, R. Janoutova^g, R. Paugam^h, F. De Boissieuⁱ, A. Kallel^j, Z. Malenovský^b, J.-P. Gastellu-Etchegorry^a

^a CESBIO, Toulouse, France

^b University of Bonn, Bonn, Germany

^c University of Jilin, Jilin, China

^d CAS, Beijing, China

^e Hongkong Polytechnic University, China

^f NASA, USA

^g CzechGlobe, Czech Republic

^h CERTEC, UPC, Spain

ⁱ TETIS, Montpellier, France

^j CRNS, ATMS, Sfax, Tunisia

Outline

- 1. Introduction to DART model**
- 2. DART applications**
- 3. Conclusion**



Outline

1. Introduction to DART model

2. DART applications

3. Conclusion





DART model: an overview

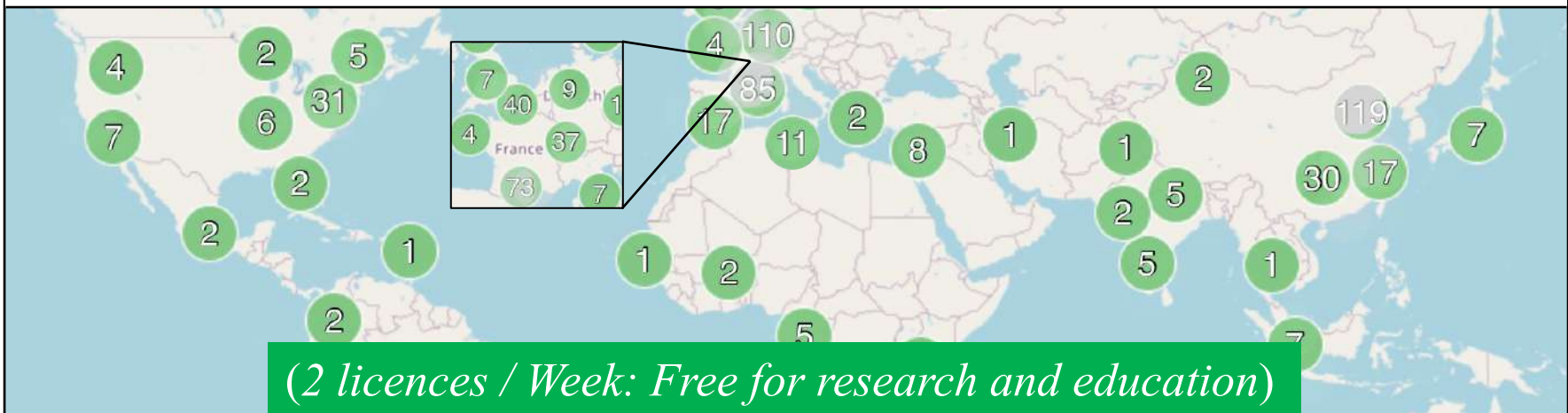
History: developed in CESBIO since 1992 by 10 scientists. Patented in 2003

Code: 500,000 lines C++ (RT), 100,000 lines Java (GUI) + Python (Tools)

Accuracy (relative difference ε , RMSE) assessed with:

- Monte Carlo models (RAMI-III experiment): $\varepsilon_{\rho} \leq 1\%$ ([Widlowski et al., 2007](#))
- Measurements: $\varepsilon_{\rho} \leq 2.5\%$ ([Landier et al., 2018](#)), $RMSE_{T_B} < 2K$ ([Sobrino et al., 2011](#))

589 DART licences: Universities, Research centres (CNES, ESA, ...)





DART Team (CESBIO)

Jean-Philippe Gastellu-Etchegorry

Professor (UT3)
Scientific leader



Nicolas Lauret

Dr, Engineer (CNRS)
Lead Developer



Science

Yingjie Wang
ATER (UT3)

Atm., polarization

Zhijun Zhen

Lecturer (Univ. Jilin)

Inversion

Paul Boitard

PhD (UT3)

Biosphere processes

Romain Demoulin

PhD (UT3)

Vegetation

Ameni Mkaouar

Post-Doc (NASA)

LiDAR

Huaan Jin

Assoc. Prof (CAS)

Vegetation

Computer science

Jordan Guilleux

Engineer (CNRS)

Interfaces, databases, ...

Eric Chavanon

Engineer (UT3)

Compilation, Scientific tools, ...

Outside CESBIO:

Z. Malenovsky, O. Regaieg, T. Nguyen (Univ. Bonn, Germany): SIF, TIR, RB.

A. Kallel (CRNS, Tunisia): Monte Carlo

T. Yin (HPU, China): Photogrammetry, LiDAR

R. Paugam (UCP, Spain): Fire

TETIS (Montpellier): F. De Boissieu, J.-B. Feret, S. Durrieu

Pytools4dart: <https://gitlab.com/pytools4dart>



Development of DART model

Since 1992 Radiative transfer: Adaptation of DOM

DART-FT

(Optical and thermal radiometer images and radiative budget, with SIF)

Since 2010 Radiative transfer: forward MC

DART-RC

(LiDAR: waveform, point cloud, photon counting)

Since 2018 Radiative transfer: bidirectional MC

DART-Lux

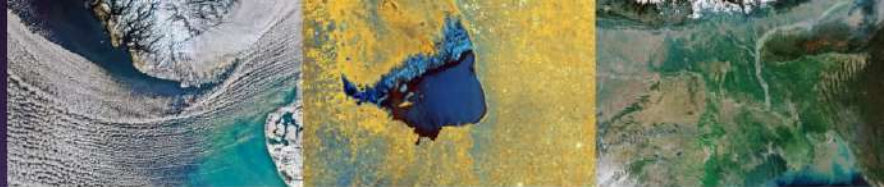
(Optical and thermal radiometer images, LiDAR and radiative budget, with SIF and polarization)

Since 2020 Energy balance

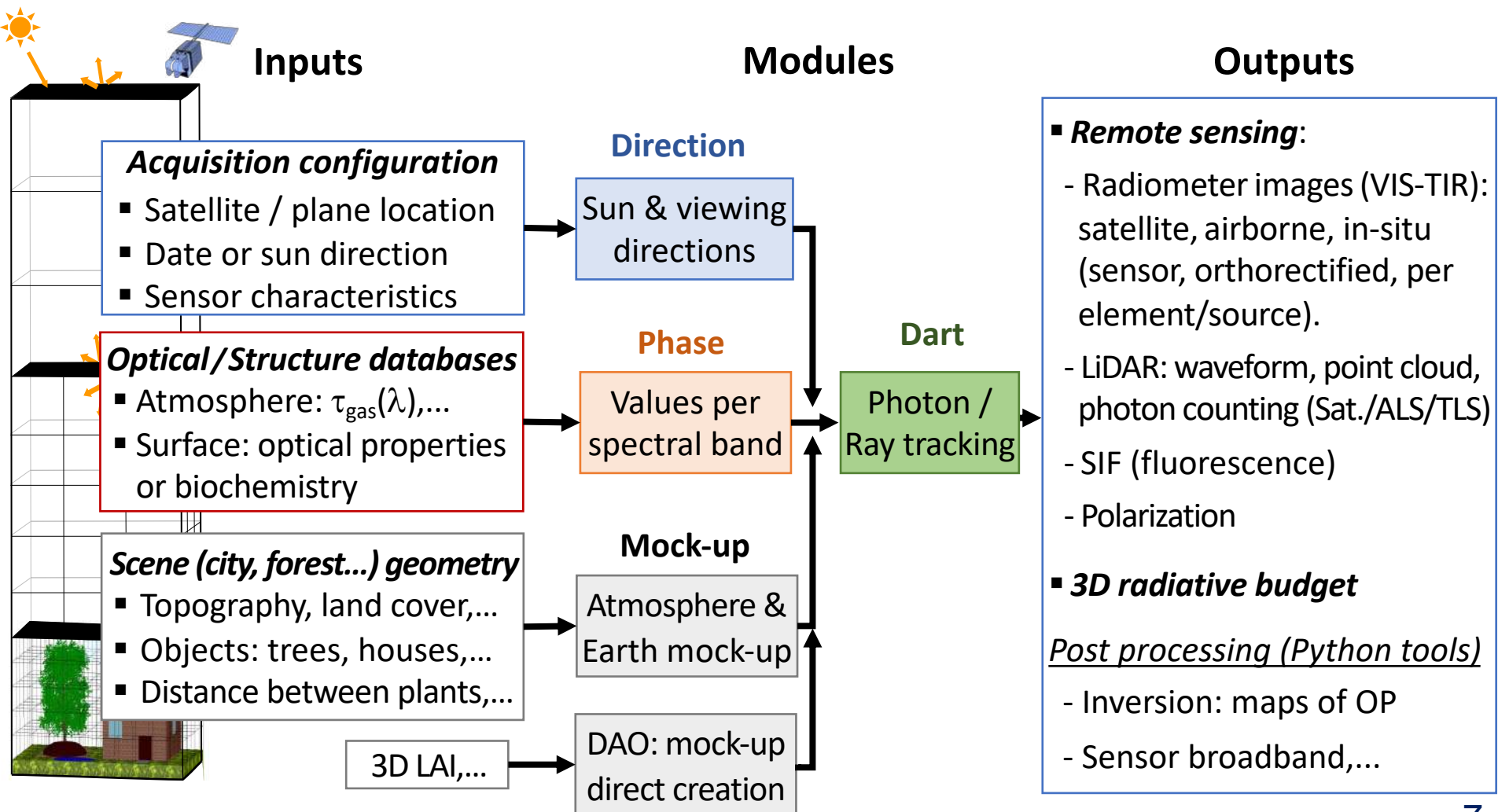
DART-EB

(Heat flux, temperature, evapotranspiration)

More than 30 years constantly improvement for land surface studies

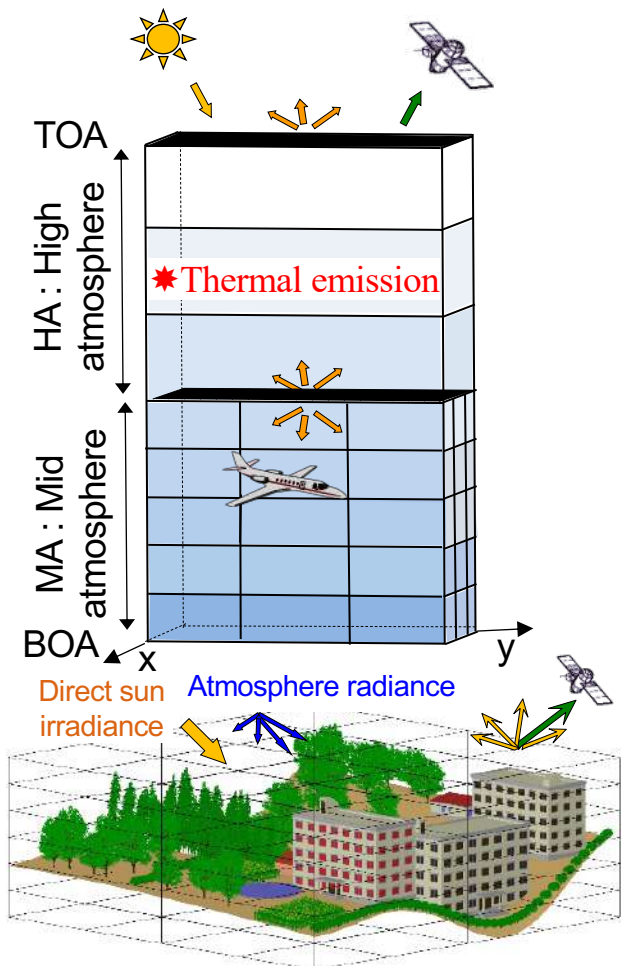


DART: an overview





DART-FT (since 1992)

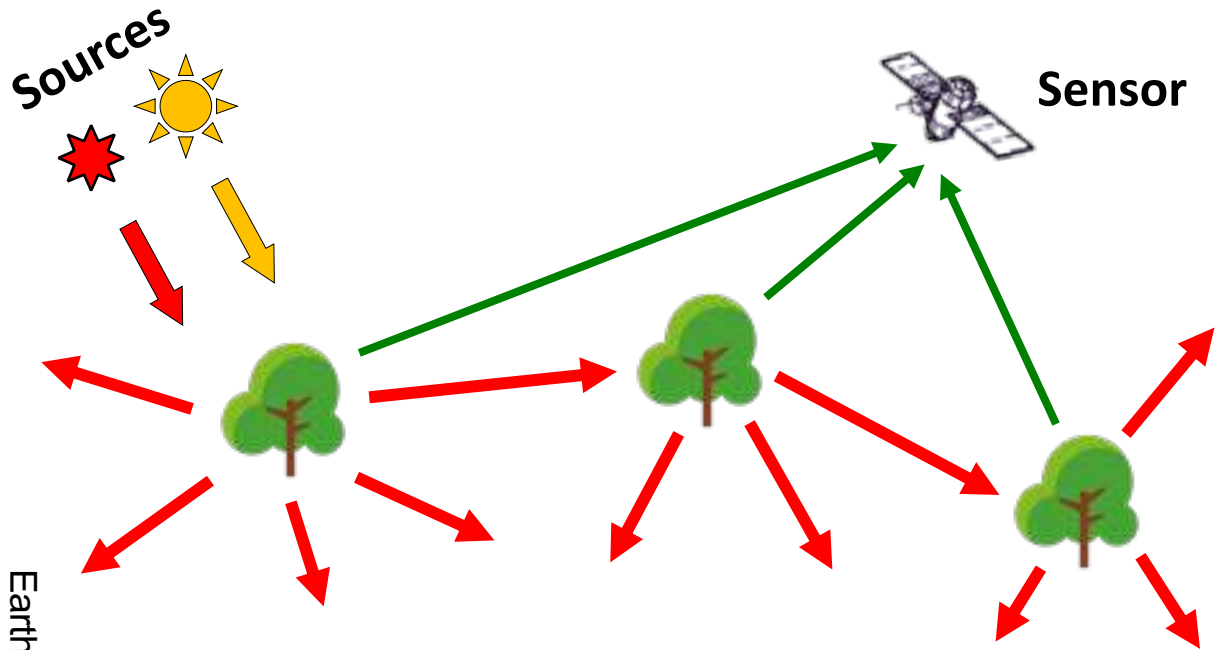


(Gastellu-Etchegorry et al., 1996)

Scene: Array of turbid / fluid voxels + facets

RT modelling: - Discrete Ordinate Method (DOM)

- **Iterative:** rays intercepted at iteration n are scattered at iteration $n + 1$



Products: VIS / TIR radiometer images & RB

RB (radiative budget) \Rightarrow simulation of photosynthesis, evapotranspiration, ...

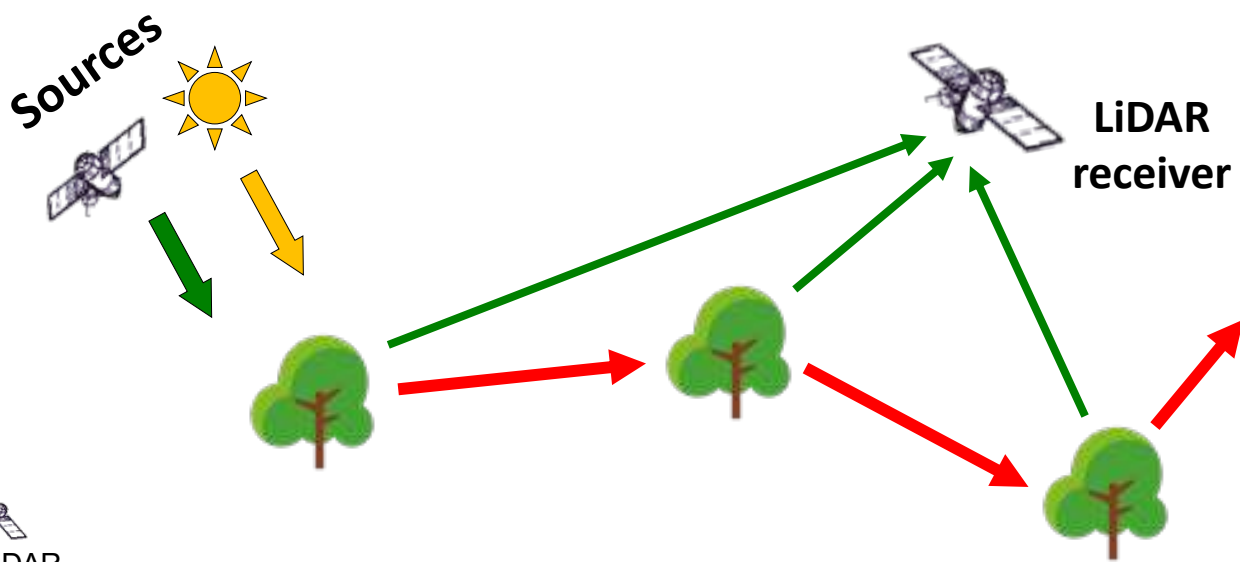
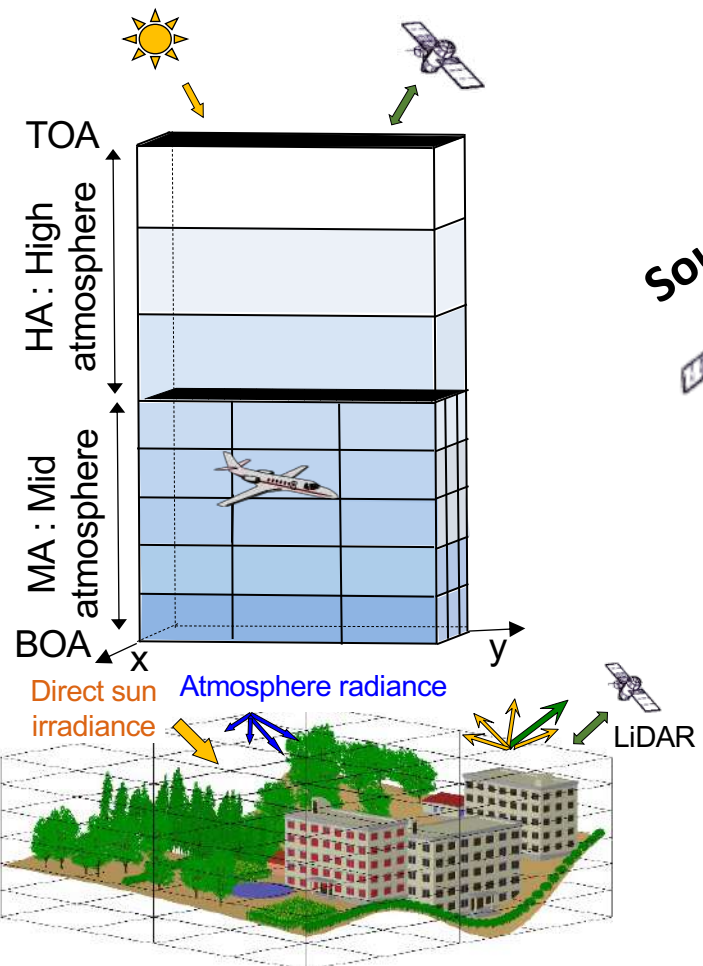


DART-RC (since 2010)

Scene: Array of turbid / fluid voxels + facets

RT modelling: - Forward Monte Carlo

- **Secondary ray:** a ray is sent to sensor after each interaction.



Products: Satellite, ALS and TLS LiDAR

(Yin et al., 2015)



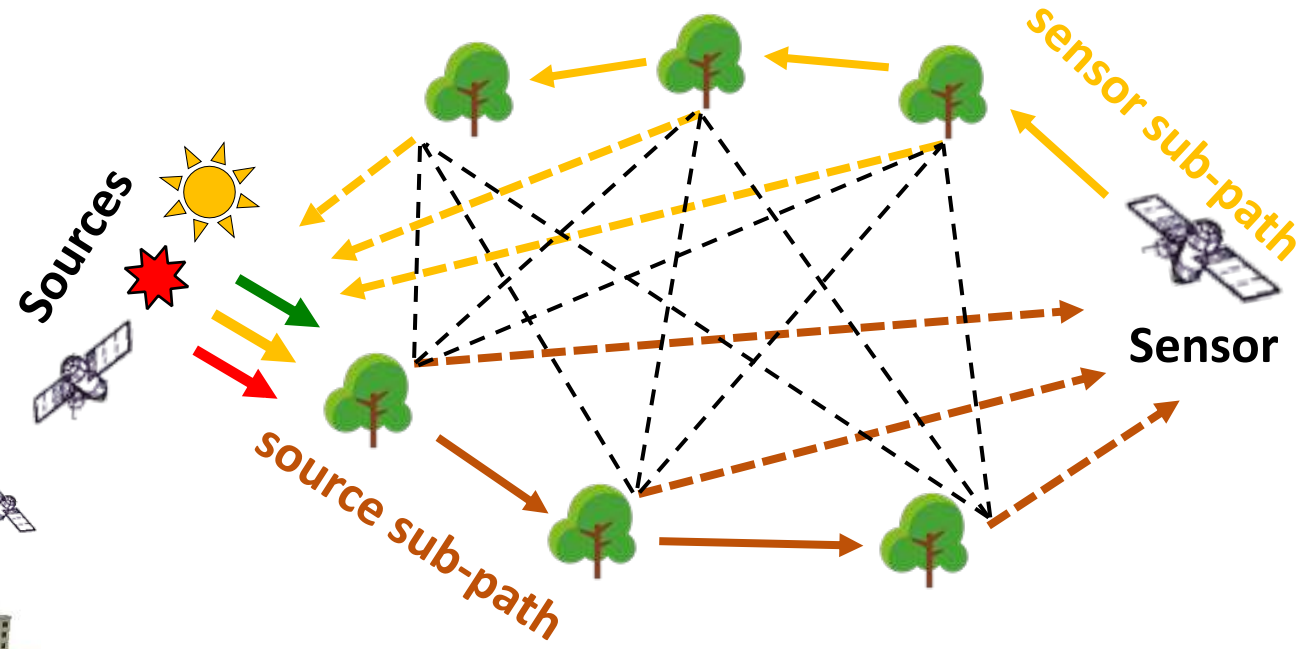
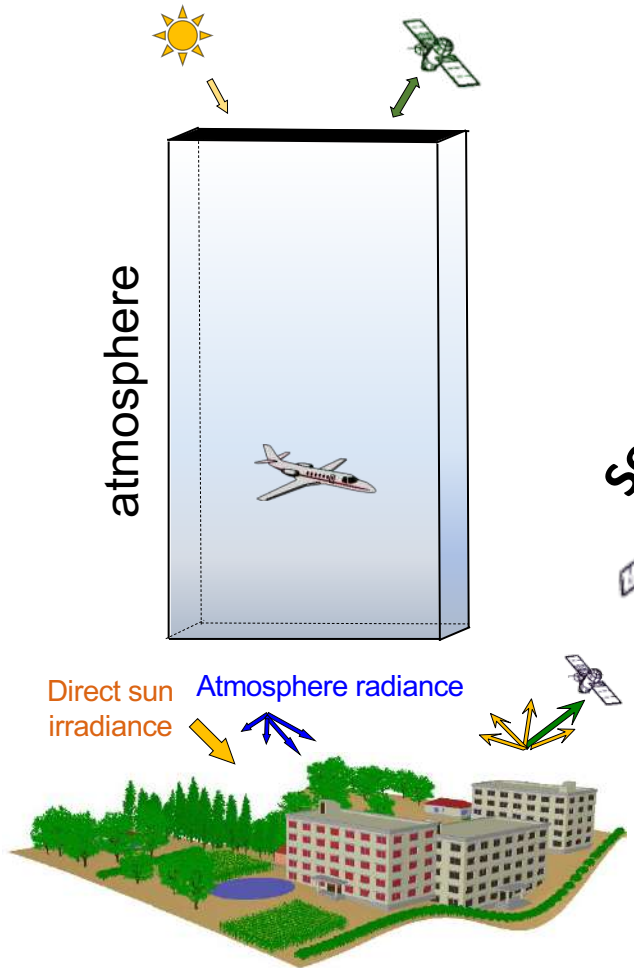
DART-Lux (since 2018)

Scene: Turbid / fluid volumes, facets; no voxel

RT modelling: - Bi-directional Monte Carlo

⇒ Computer time & RAM reduced by > 100 !

Rays run over most probable "Source - Sensor" paths



Products: VIS / TIR radiometer images, RB and LiDAR



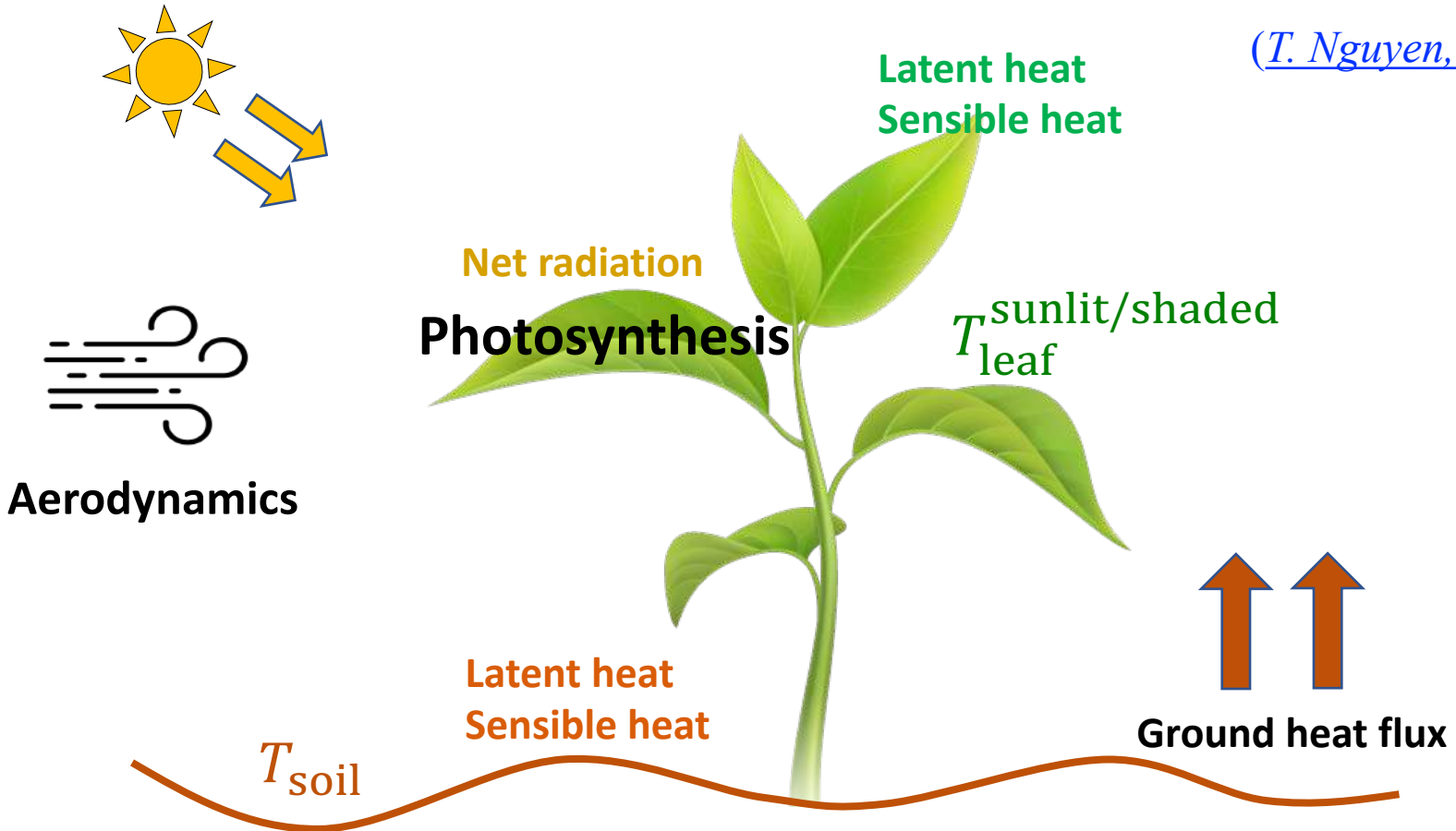
DART-EB (since 2020)

Scene: 1D (3D scene modelling is on-going)

EB modelling: Photosynthesis, heat flux, turbulence, fluorescence, evapotranspiration, *etc.* + RT (DART)

Products: Vertical heat flux, temperature profile, ...

(T. Nguyen, 2022)



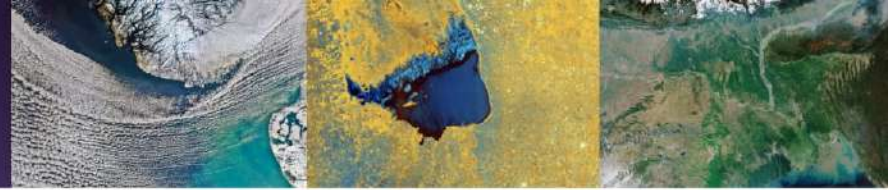
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DART applications

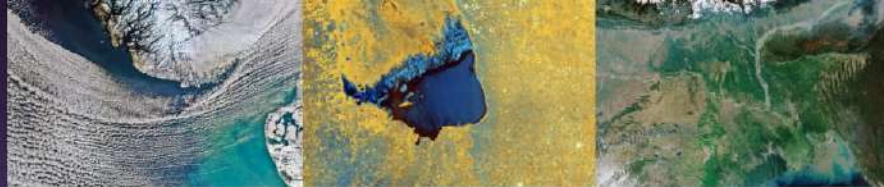
Research works with indication of projects:

- Vegetation functioning (hyperspectral): RedEdge, HyPOS (ESA)
- Urban radiative budget: H2020 UrbanFluxes, ERC Urbisphere (EU)
- AI algorithms: DIAPOS (CNES)
- Normalization of satellite images: COPA (ESA)

Preparing satellite missions:

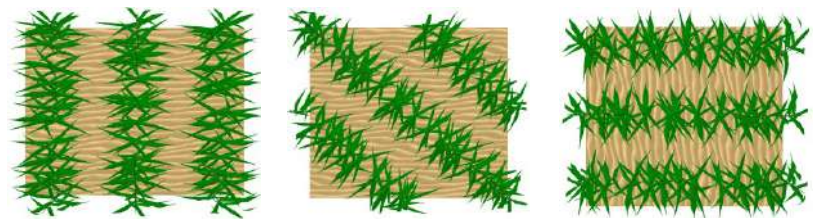
- Sentinel-2 NG (ESA): trade-off "Several satellites vs. Large FOV"
- TRISHNA (CNES-ISRO), LSTM (ESA): optimal TIR bands, DART scenes used as references, correction of directional effects in TIR images
- Surface Topography and Vegetation (STV: NASA): LiDAR, stereo, ...

Community code certification (French National Research Centre, **CNRS**): to facilitate scientific collaboration on research domains using DART (**Submitted**)

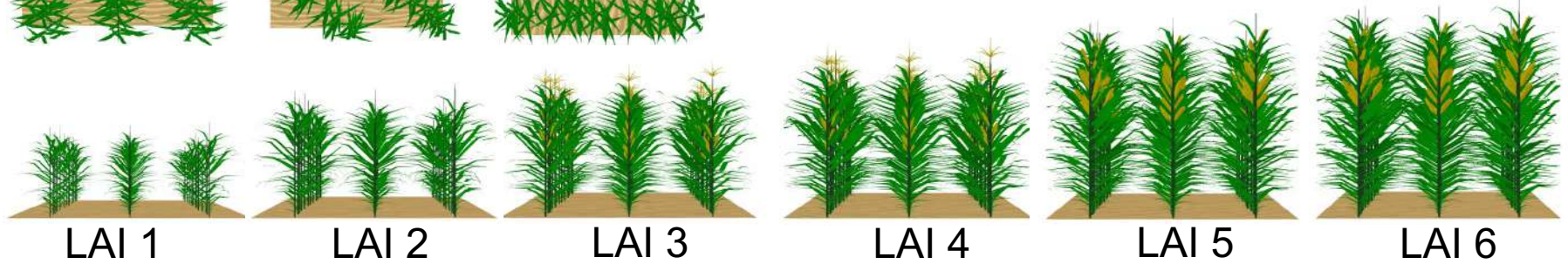


Remote sensing images

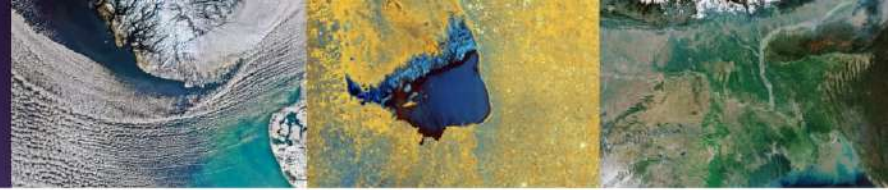
Impact of vegetation growth stage and field orientations on satellite signal and radiative budget. [Projects LSTM & S2 NG \(ESA\)](#)



**Maize at 6 growth stages (LAI, height...)
3 orientations, 3 ground reflectances, ...**



	Maize stage (BBCH scale)					
Plant height	0.92	1.28	1.55	1.79	2.04	2.24
Number of leaves	14	14	19	19	22	22
LAI	1	2	3	4	5	6
Leaf max. area	0.105	0.21	0.315	0.42	0.525	0.63

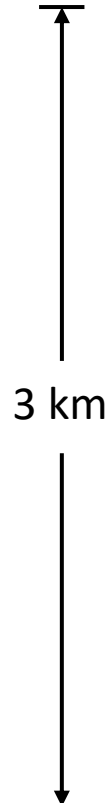


Remote sensing images

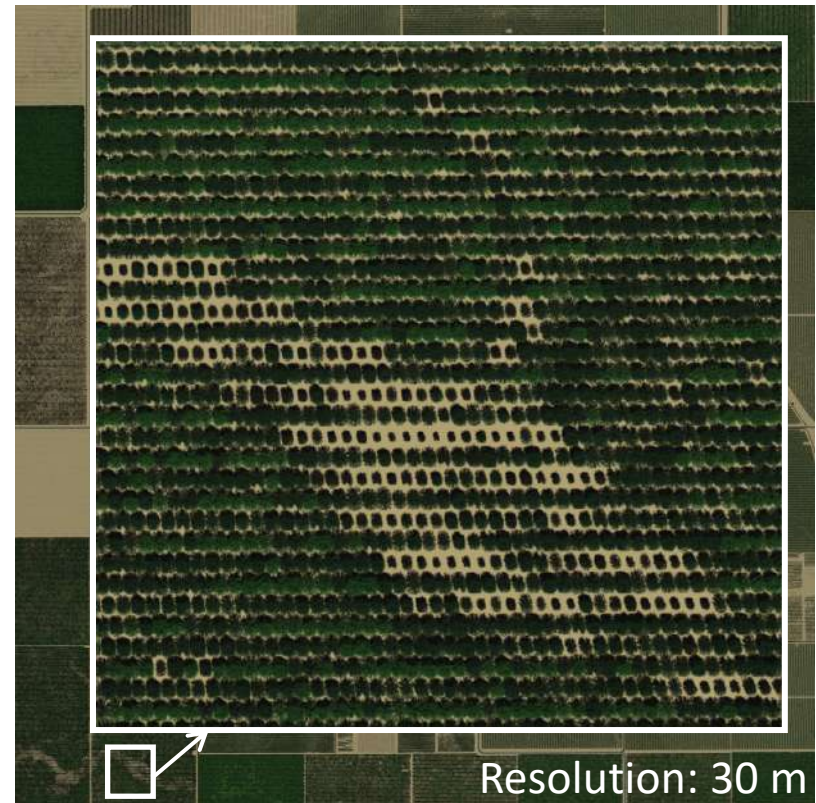
DART scenes used as ref. \Leftarrow High resolution images. Project S2 NG (ESA)
(Gastellu-Etchegorry et al., 2022)



Google map (Sept. 2018)

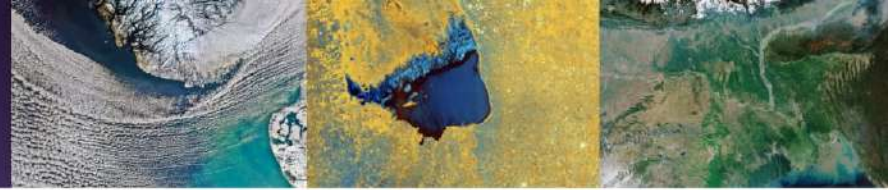


3 km



Resolution: 30 m

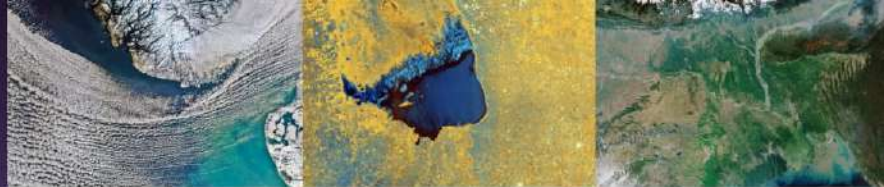
DART Simulation



Remote sensing images

Urban short wave and long wave radiative budget. Project Suabe (Belgium)

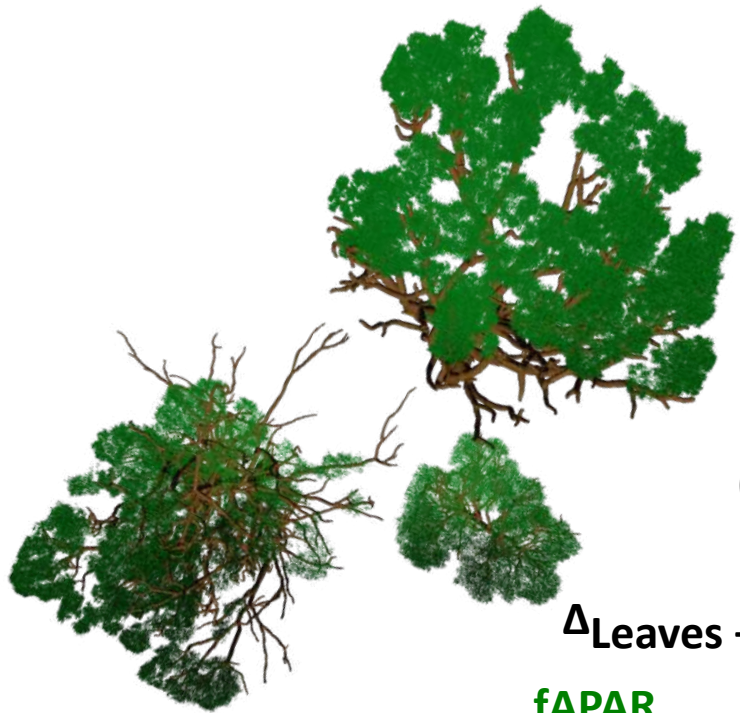
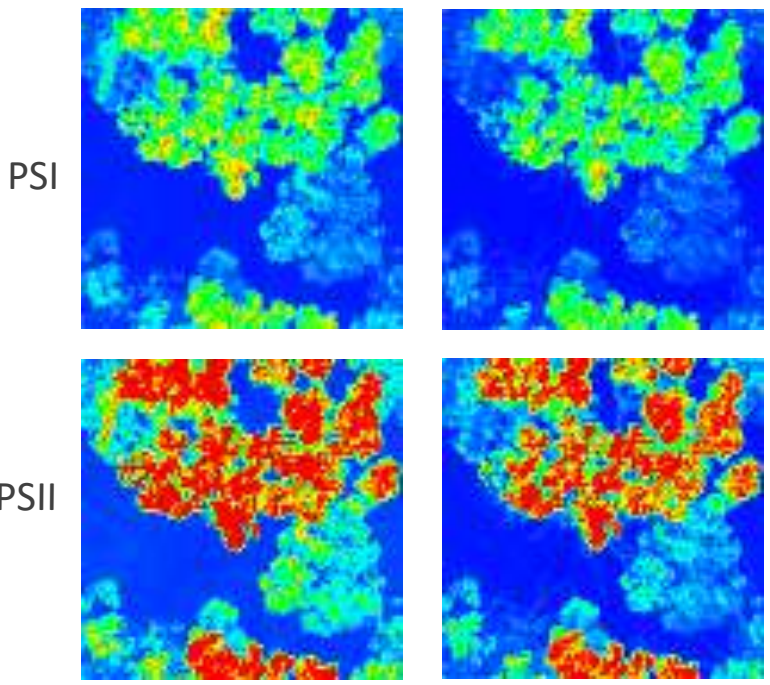




Chlorophyll fluorescence

Impact of wood on SIF emission and observation at 740nm: Eucalyptus forest

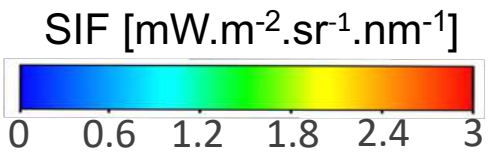
Only leaves Leaves + wood



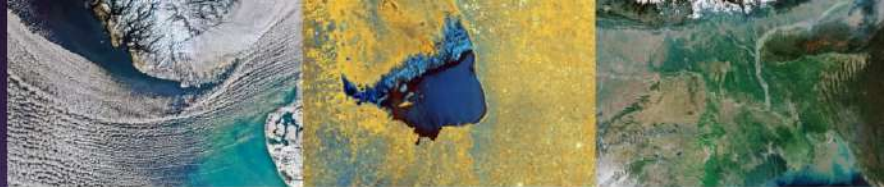
Eucalyptus trees (DART simulation)

$\Delta_{\text{Leaves} - \text{Leaves+Wood}} \%$

fAPAR _{green}	17% ↓
SIF emitted	17% ↓
SIF exitance	24% ↓



(Malenovsky et al., 2021)

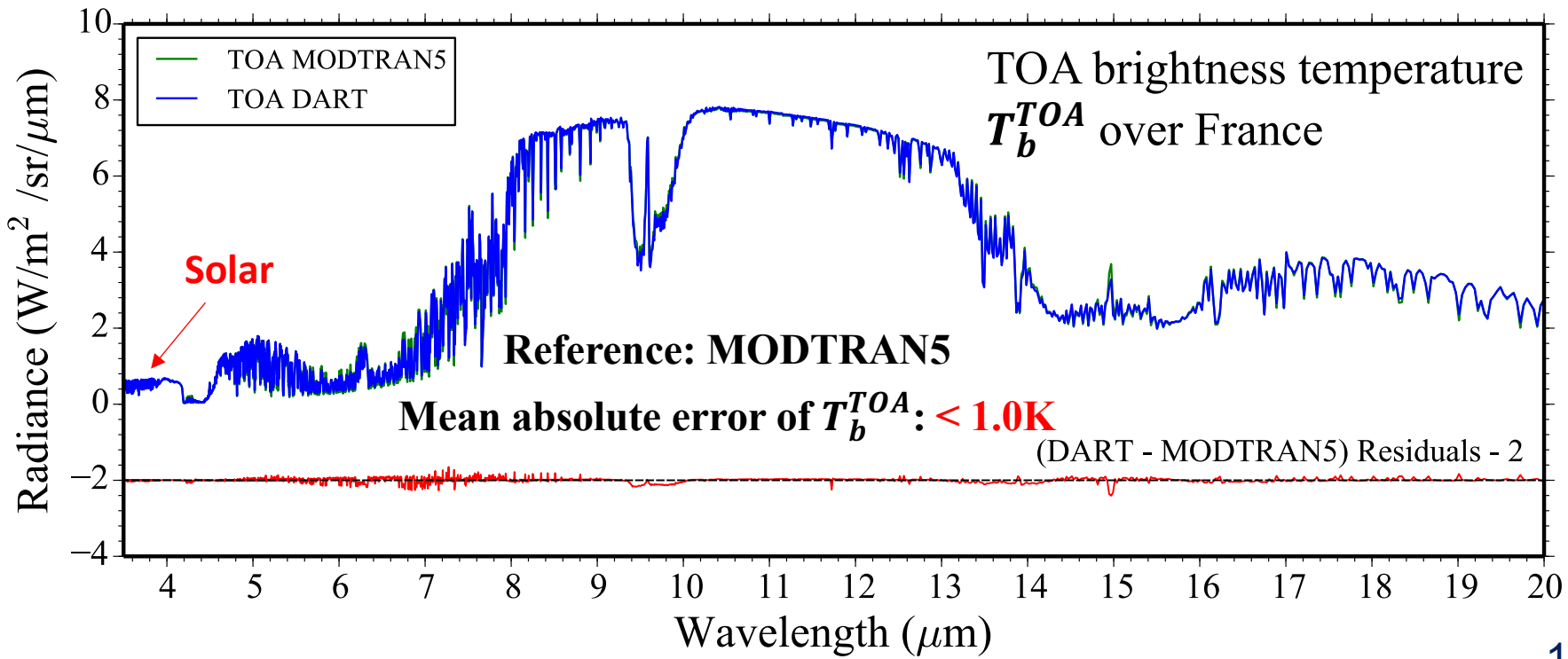


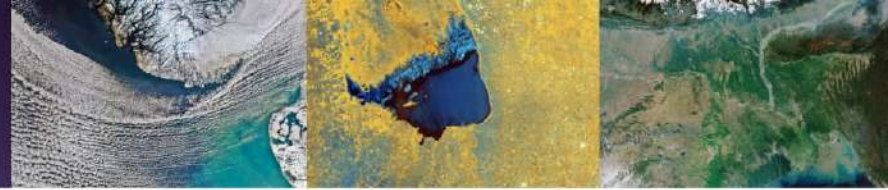
Atmosphere

ECMWF: vertical profile of gases and temperature. **DART:** atmospheric RT



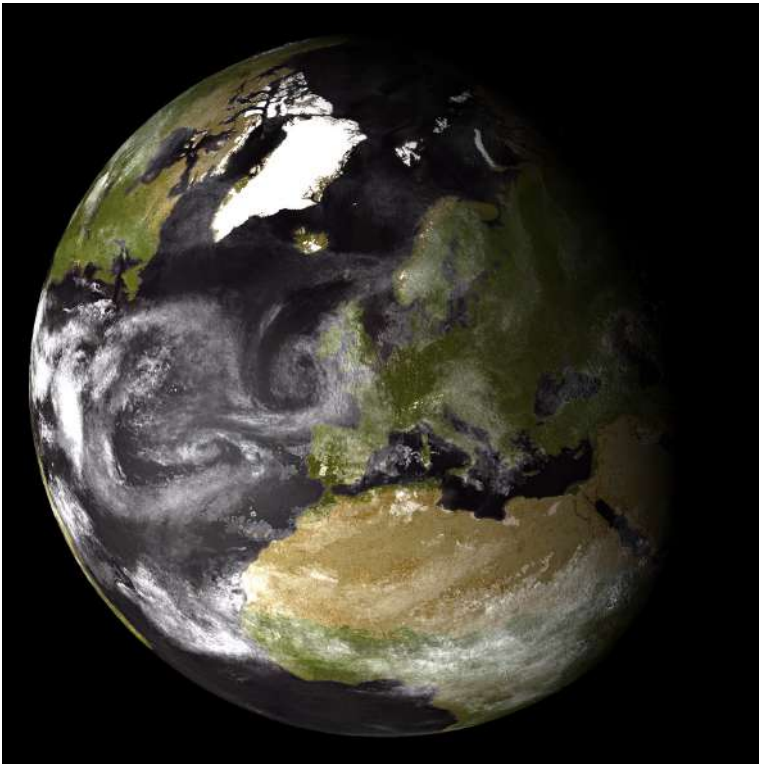
(Wang et al., 2020)



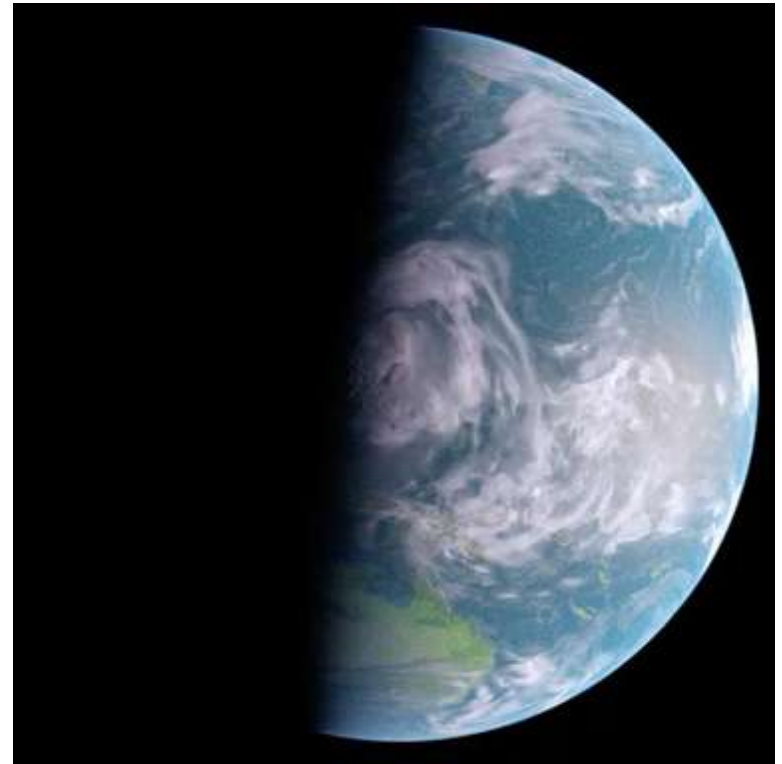


Atmosphere

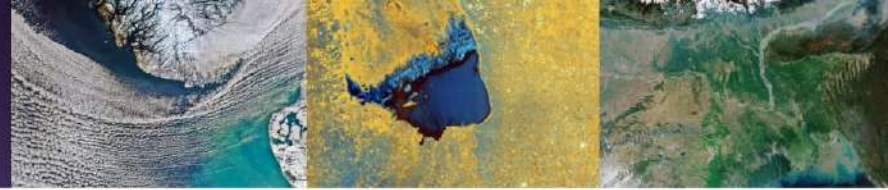
Data sources: Earth DEM (GEBCO 2022), Land cover (EarthEnv), Clouds cover (Copernicus EUMETSAT), Atmosphere profile (AFGL: USSTD76)



No Atmosphere



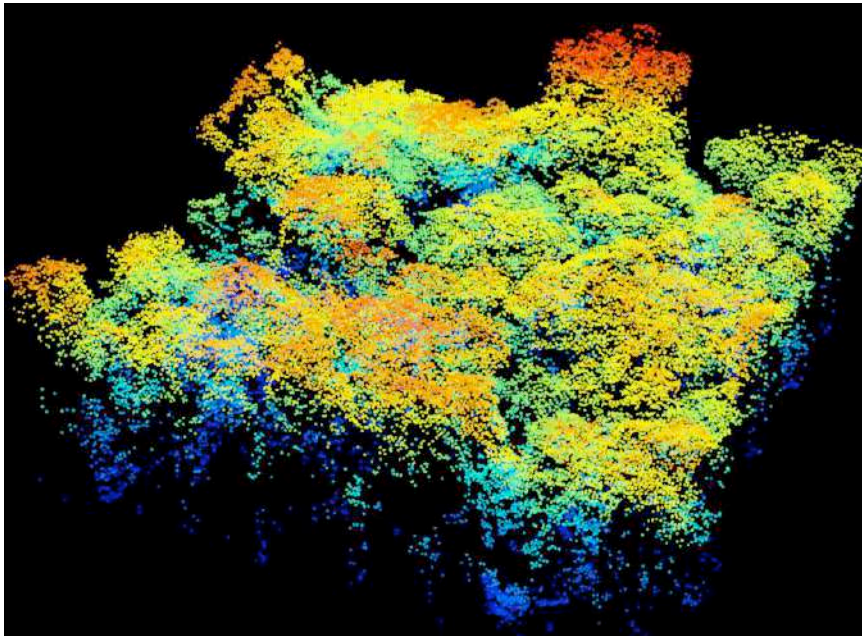
Spherical Atmosphere (E. Chavanon) 19



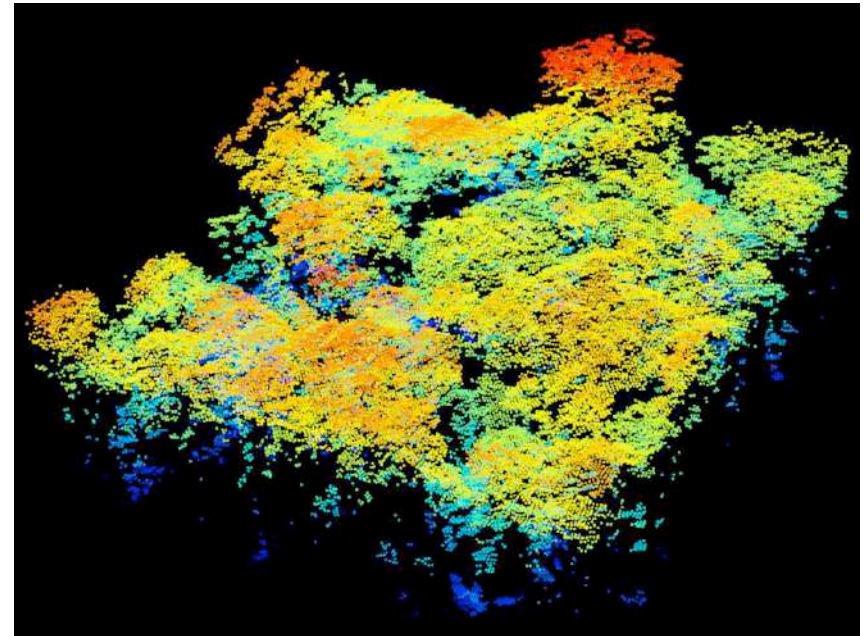
LiDAR

Simulation of airborne and satellite LiDAR signals of tropical forest (Paracou, Guyana) in order to prepare the satellite mission LEAF (LiDAR)

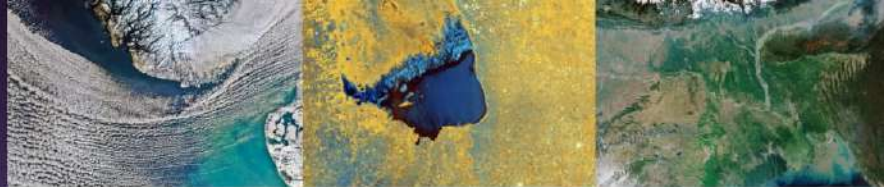
(Durrieu et al., 2019)



Airborne **Riegl LMS-Q780**

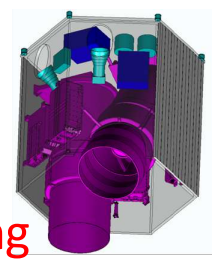


DART: Measurements \Rightarrow optical properties
TLS \Rightarrow 3D architecture



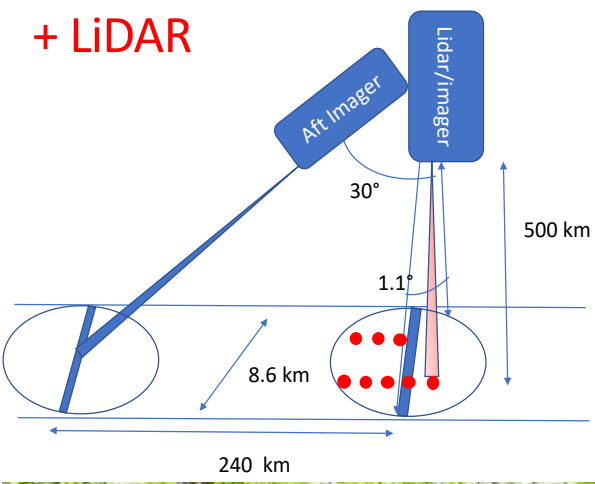
LiDAR

Design next generation instrument \Rightarrow HR global topography. Project STV (NASA)

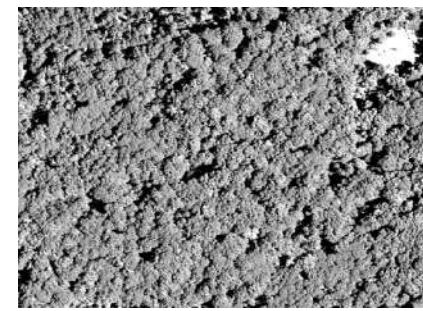


DART gives images and (x, y, z) coordinates of scene elements
 \Rightarrow NASA uses DART to define the optimal satellite configuration

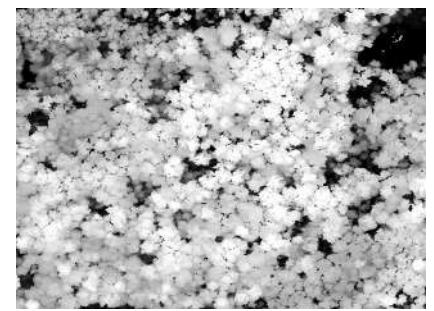
Stereo imaging
 + LiDAR



DART: 400x400m SERC forest



DART: nadir (15/06/2012, 9am)



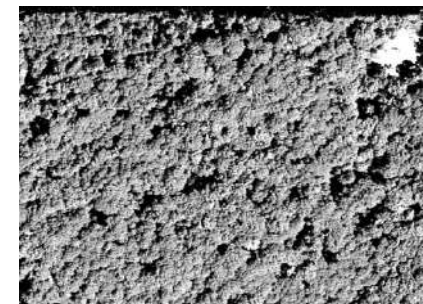
NASA G-LiHT Lidar Reference



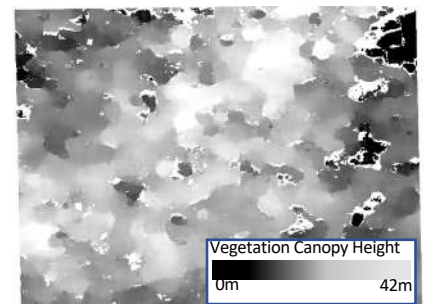
SERC forest



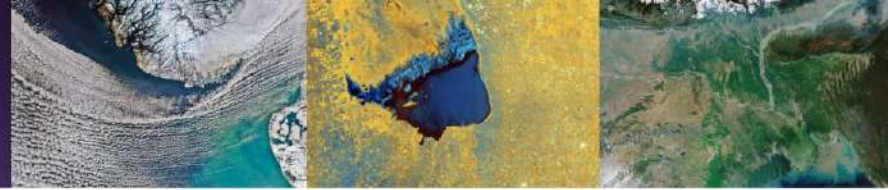
DART: scene-leaf off



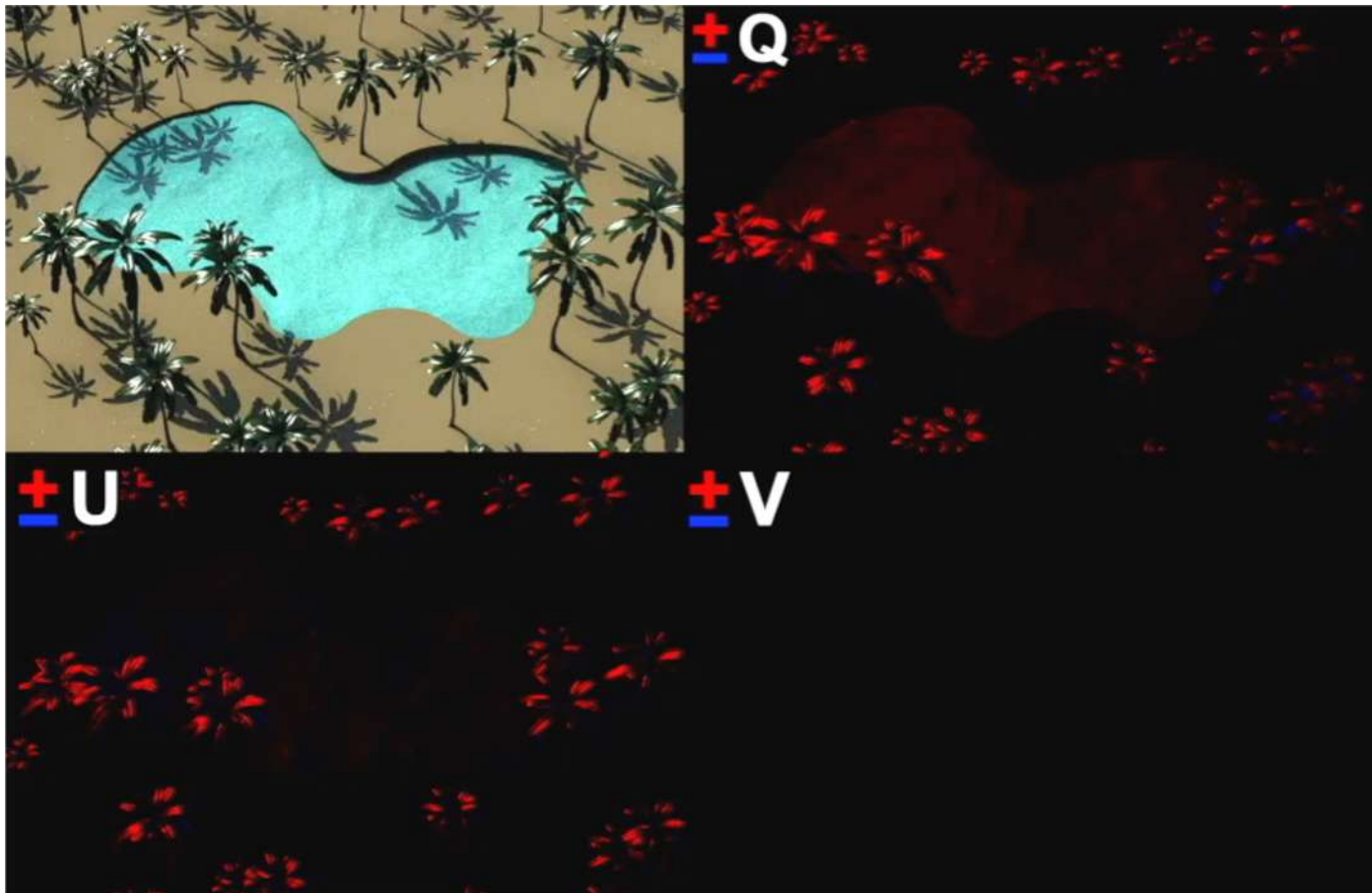
DART -20° (15/06/2012, 9am)



Canopy height from DART stereo pair



Polarization



**Specular reflection
& Polarization**

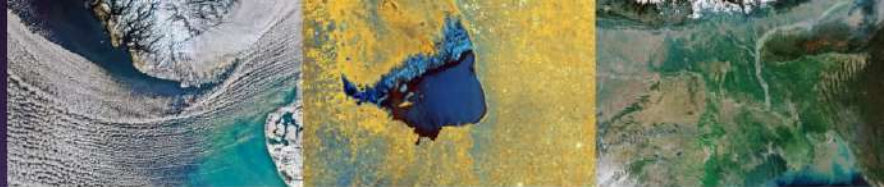
**Stokes vector:
[I, Q, U, V]**

**Future polarimetric
satellite missions:**

METOP-SG-A (ESA)

OTB-2 (NASA)


(Wang Yingjie, 2022)



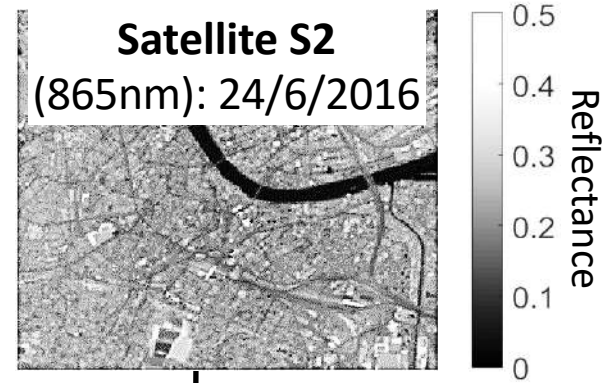
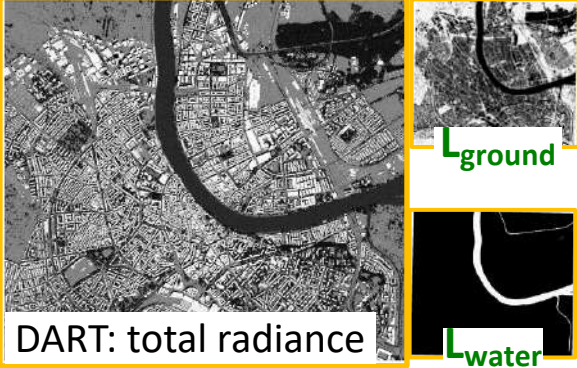
Inversion

(Zhen et al., 2021)

Urban database



+ Spatially constant **optical properties (OP)**
+ **Atmosphere** and **Satellite** configurations

DART

L_{ground}

L_{water}

DART: total radiance

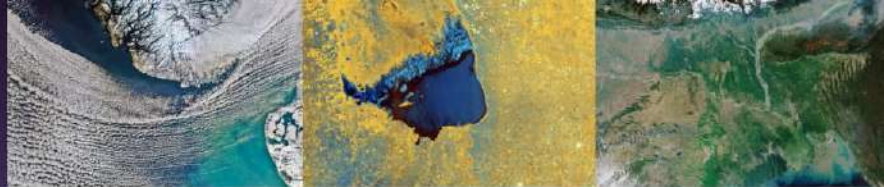
DART

DART
($\epsilon < 10^{-3}$)

New OP maps
of roofs, streets,...

Solving linear equations
using DART Jacobian

- **Satellite image: any sun, view, atmosphere**
- **Albedo & RB maps: time series, satellite driven**

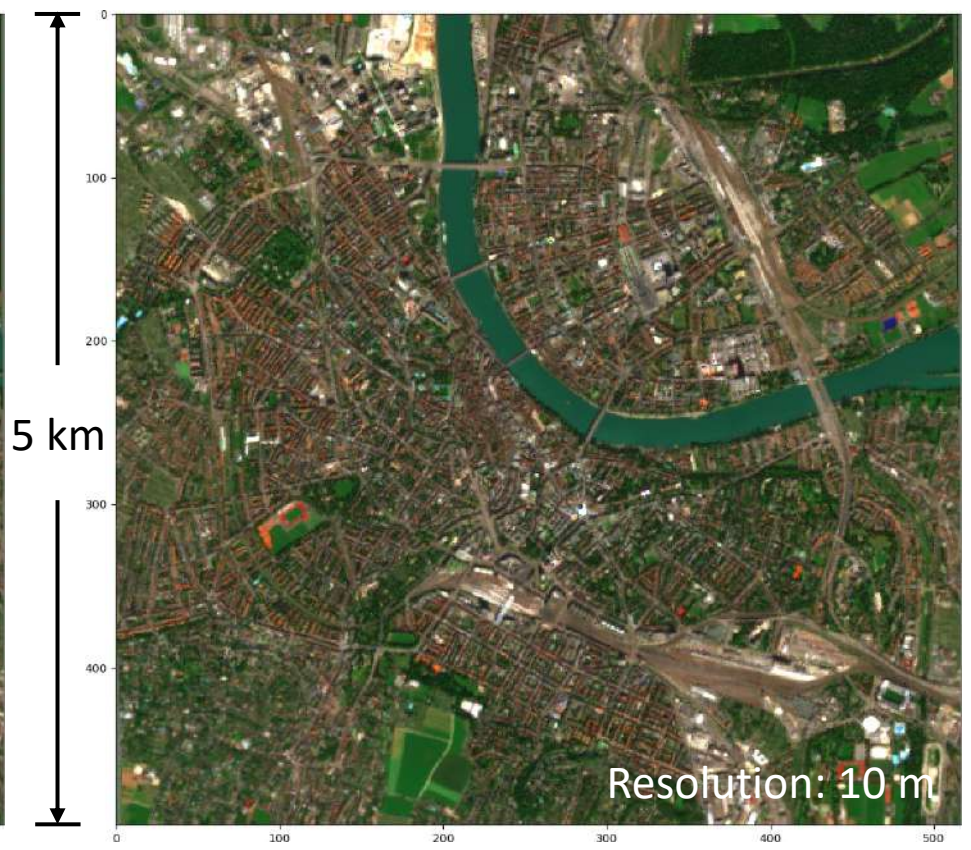


Inversion

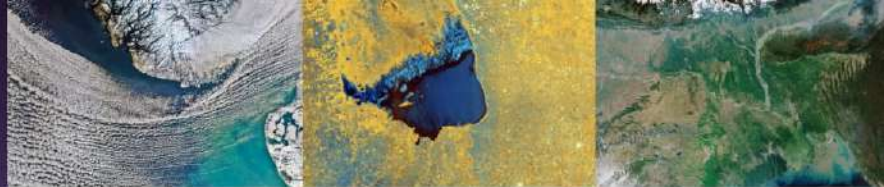
(Zhen et al., 2021)



Sentinel 2 (B2, B3, B4)

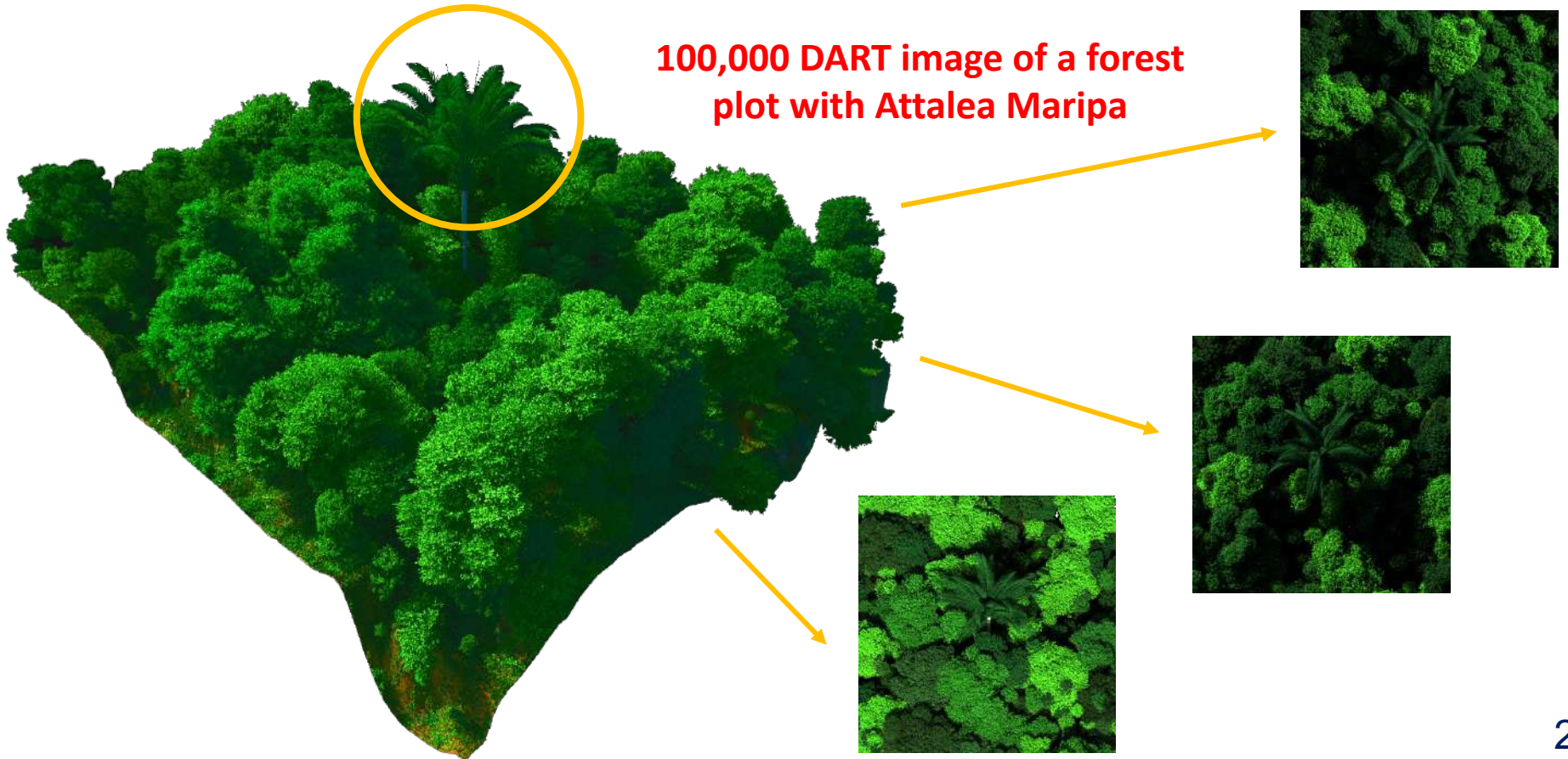


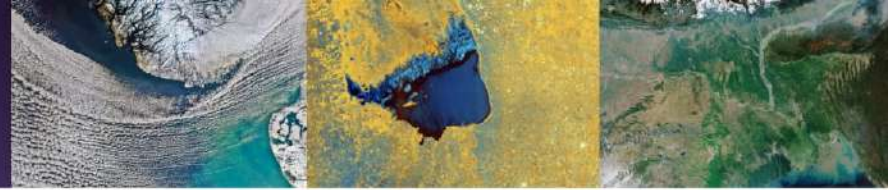
DART simulation with OP maps



Deep learning

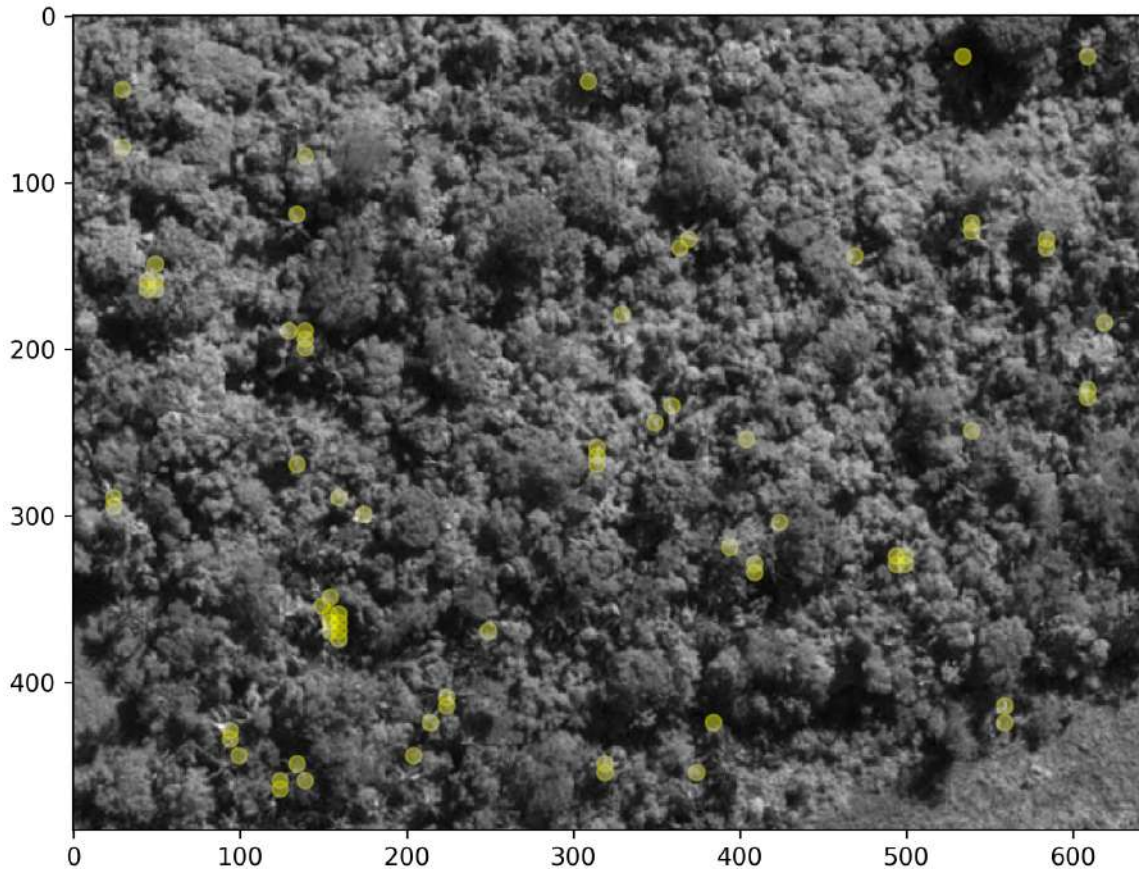
Training an **AI algorithm** to detect palm trees in high resolution satellite images. Project **DIAPOS** (CNES)





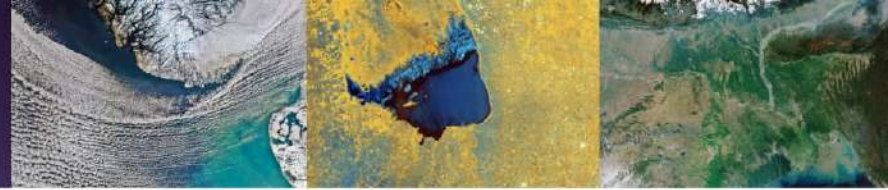
Deep learning

Preliminary evaluation: **Detection precision 75 %**



Automatic detection of palm tree in high spatial resolution satellite image.

(Data from ESPACE-DEV, IRD)

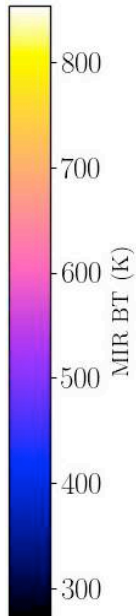
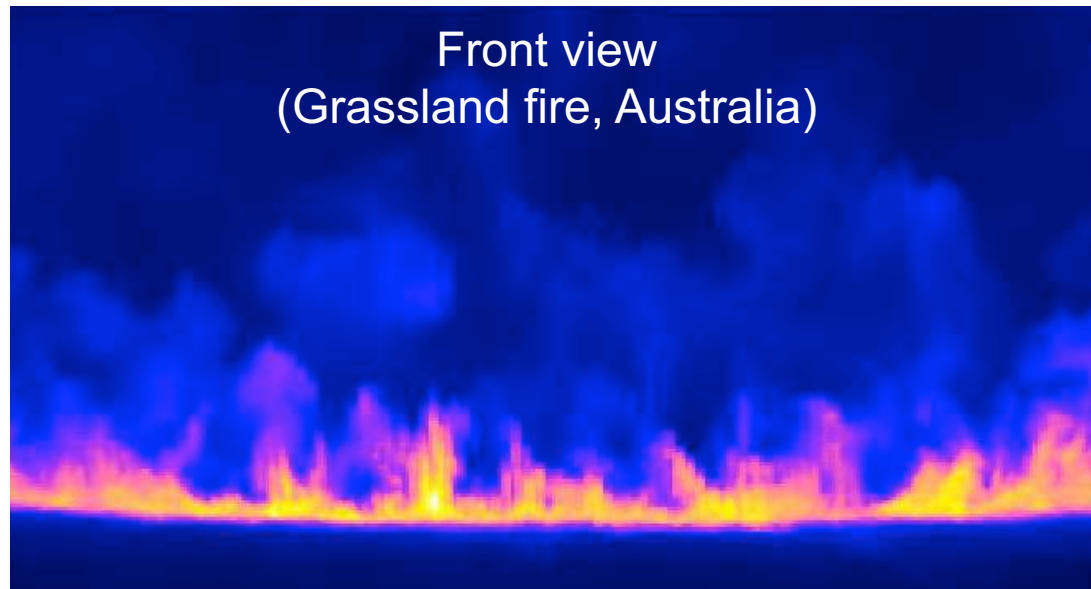
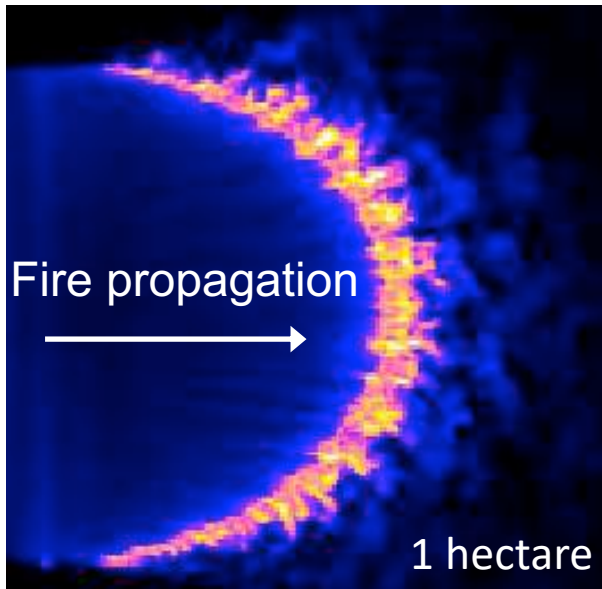


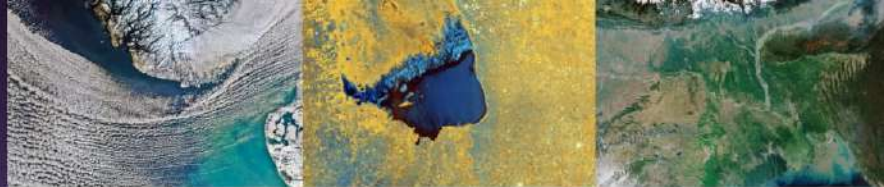
Fire

FDS (Fire Dynamics Simulator) model \Rightarrow 3D temperature distribution
3D soot/gas density

DART \Rightarrow Remote sensing observations (TIR camera, satellite)

\Rightarrow Study the **fire radiative power** from satellite observation





Urban surface temperature

SOLENE model \Rightarrow 3D energy balance (2 broad bands) \Rightarrow LST + T_{air}

DART model \Rightarrow hyperspectral RTM (more accurate RB) \Rightarrow RS observations

Impact of urban surface heterogeneity on LST estimation from TIR satellites
(**TRISHNA, LSTM**)

Brightness temperature at 4 view zenith (vz) angles

LST – SOLENE microclimate



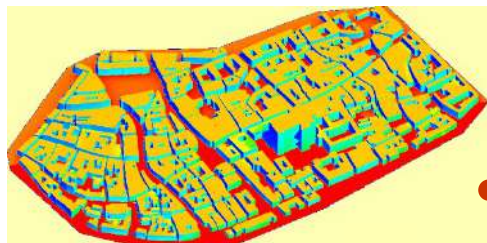
Strasbourg, cathedral district,
15th June 2021, 9h UTC



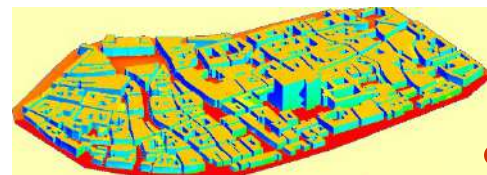
$vz = 15^\circ$



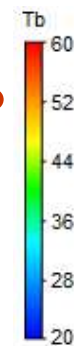
$vz = 30^\circ$



$vz = 45^\circ$



$vz = 60^\circ$



● Sun position

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Conclusion

DART:

- 1) **Efficient & accurate** for simulating 3D surface observations (TIR, SIF,...) and radiative budget at various spatial and time scales
- 2) Adapted to a **wide range of applications**: inversion of satellite data, preparation of satellite mission, AI, fire, ...
- 3) Easy to chain with **process models** (SOLENE, SCOPE,...)

On-going work

- (1) 3D modelling of energy balance of vegetation (DART-EB)
(**temperature distribution** \Rightarrow TIR obs., RB)
- (2) Differentiable radiative transfer (Jacobian)
(**uncertainties, better inversion**)
- (3) Technical: georeference, NetCDF format, GPU, ...
(**preparing satellite missions, massive simulation**)



THANK YOU

DART is freely available for research and education.

Contact:

jean-philippe.gastellu@iut-tlse3.fr

nicolas.lauret@univ-tlse3.fr

yingjie.wang@univ-tlse3.fr

(J.-P. Gastellu-Etchegorry)

(Nicolas Lauret)

(Yingjie Wang)