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# RETRIEVAL OF LEAF AND DRY MATTER CONTENTS USING RADIATIVE TRANSFER MODELING AND NIR IMAGING SPECTROSCOPY

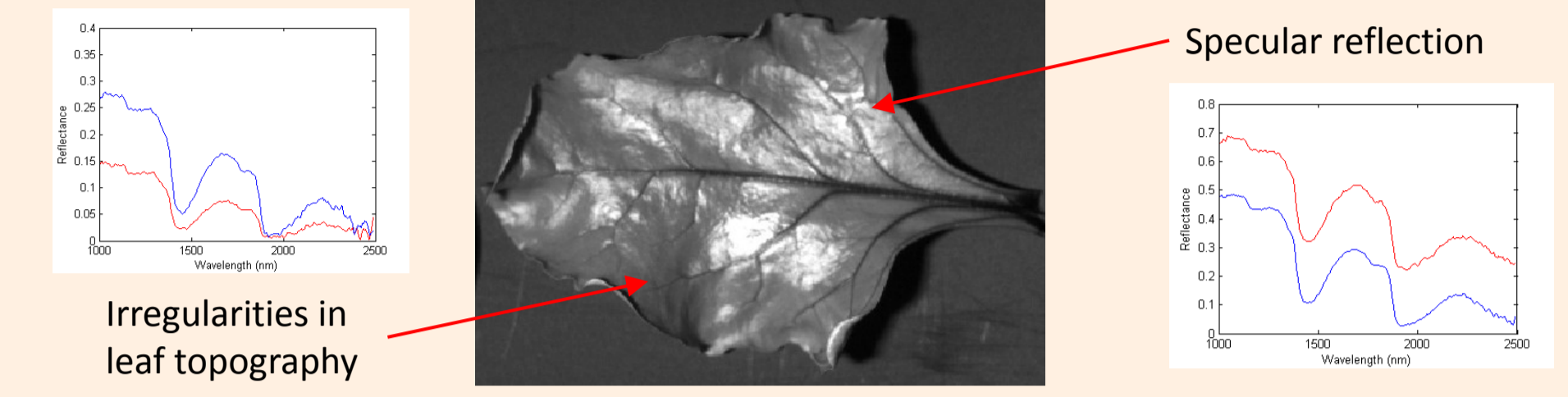


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## Context

- Imaging spectroscopy**, a powerful tool for spatially characterizing vegetation:
  - ✓ Identification of plant pigments (Blackburn, 2007),
  - ✓ Detection of freezing stress (Nicotra et al, 2004),
  - ✓ Detection of leaf diseases (Malhein et al, 2013)...
- The NIR/SWIR range** (1000-2500 nm), a relevant spectral range for retrieving leaf water content and dry matter content (strong absorption/scattering features).
- Usual methods for retrieving leaf parameters are based either on**
  - ✓ the targeted biochemical variables (statistically-based methods, machine learning algorithms),
  - ✓ the radiometric measurement (physically-based methods through the inversion of radiative transfer models)
- No radiative transfer models available for close-range imaging spectroscopy !**
- Objective:** Develop a leaf radiative transfer model for close-range imaging spectroscopy by adapting an existing leaf directional-hemispherical reflectance model.



## Theory – The COSINE model

- Hypothesis:** a close bidirectional approximation can be obtained under specific experimental and instrumental lab conditions, i.e.,
  - ✓ Collimated light beam,
  - ✓ Small sensor FOV (imaging spectroscopy: ok)
- What physical quantity is retrieved at the pixel level ?**
  - ✓ Hyperspectral measurement  $\rightarrow$  Bidirectional Reflectance Factor (BRF) ?  

$$R(\theta_s, \theta_v, \varphi_v) \approx \frac{L_r(\theta_s, \theta_v, \varphi_v)}{L_r^0(\theta_s)}$$
 Radiance measured on the leaf / Radiance measured on the reference surface
  - ✓ More realistically: hyperspectral measurement  $\rightarrow$  pseudo-BRF  

$$\tilde{R}_{hs}(\theta_s, \theta_i; \theta_v, \varphi_v) \approx \frac{L_r(\theta_s, \theta_i, \theta_v, \varphi_v)}{L_r^0(\theta_s)}$$
 Light incident angle = f (illumination angle, leaf topography) / Illumination zenith angle
  - ✓ It can be shown that  

$$\tilde{R}_{hs}(\theta_s, \theta_i; \theta_v, \varphi_v) \approx R(\theta_i; \theta_v, \varphi_v) \frac{\cos \theta_i}{\cos \theta_s} = R(\theta_i; \theta_v, \varphi_v) \text{ if } \theta_i = \theta_s$$
- How to model the leaf BRF as a function of leaf optical properties?**
  - ✓ **Proposed approach:** relate the leaf BRF to an existing leaf DHR model such as PROSPECT (Jacquemoud & Baret, 1990) or LIBERTY (Dawson et al, 1998).  
 Pseudo BRF  $\xrightarrow{\frac{\cos \theta_i}{\cos \theta_s}}$  BRF  $\xrightarrow{\times \pi}$  BRDF  $\xrightarrow{\Sigma}$  BRDF<sub>spec</sub> + BRDF<sub>diff</sub>  $\xrightarrow{k_{\text{lambert}}}$  DHR model  $\rightarrow$  Leaf parameters ( $C_w, C_m, \dots$ )
  - ✓ The COSINE model (C**l**ose-range S**p**ectral I**m**aging of **l**eaves)  

$$\tilde{R}_{hs}(\theta_s, \theta_i; \theta_{dhr}, b_{spec}) = \frac{\cos \theta_i}{\cos \theta_s} [\tilde{p}(\theta_{dhr}) + b_{spec}]$$
    - One known parameter ( $\theta_s$ )
    - ( $N_{dhr} + 2$ ) unknown parameters ( $\theta_{dhr}, b_{spec}, \theta_i$ )

## Materials & Methods

- Data set**
  - ✓ 5 plant species (laurel, ivy, bamboo, laurestine, holly) + sugar beet,
  - ✓ For each leaf, 3 average incident angles (30°, 10°, 50°),
  - ✓ Spectral range : 1000-2500 nm
  - ✓ Spatial resolution: 0.23 mm,
  - ✓ Destructive measurements of water content ( $C_w$ ) and dry matter content ( $C_m$ )

Parameter	Number of samples	Mean	Min/Max	Standard deviation
$C_w$ [cm]	22	0.0104	0.0049/0.0164	0.0040
$C_m$ [g.cm <sup>-2</sup> ]	22	0.0091	0.0052/0.0129	0.0023

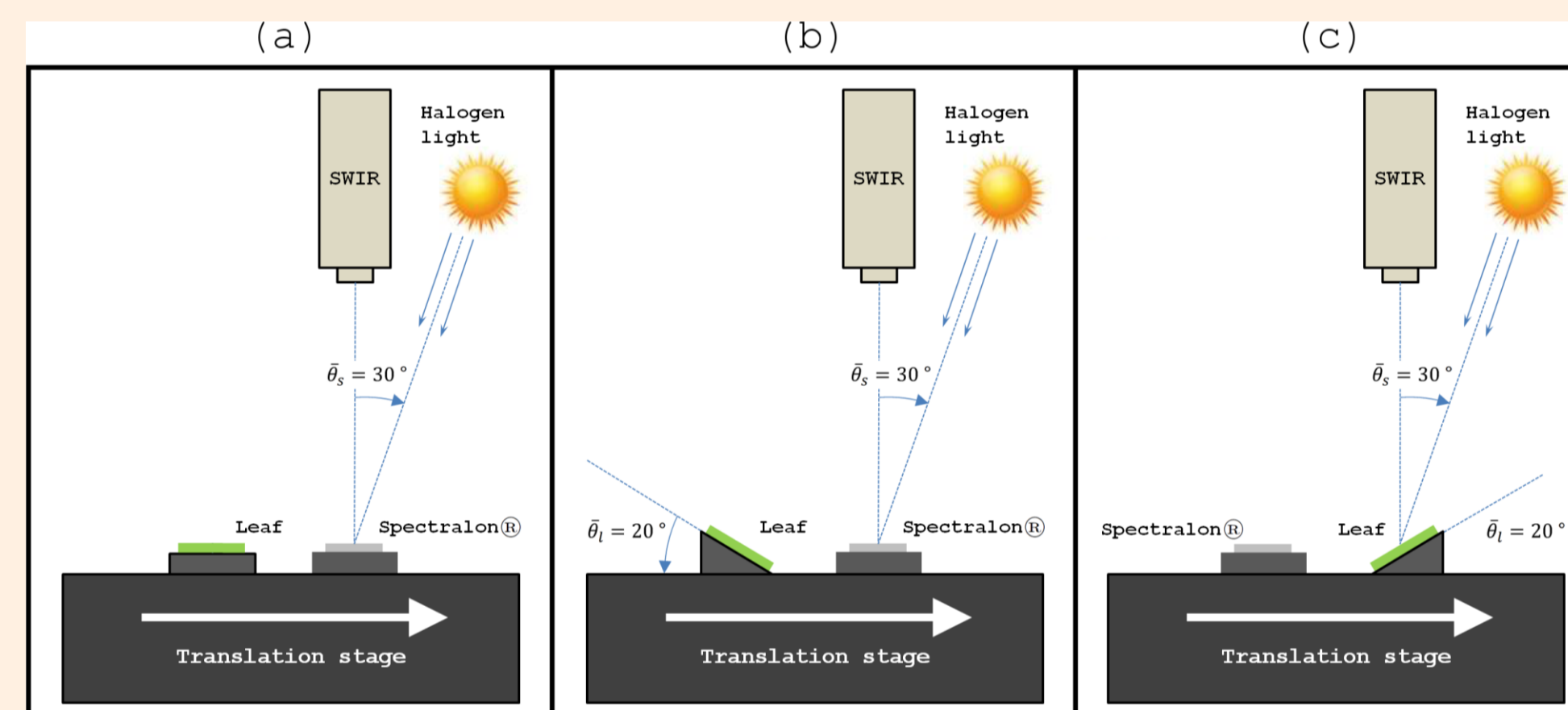


Figure: Three acquisition configurations for each leaf : (a)  $\theta_i = 30^\circ$ , (b)  $\theta_i = 10^\circ$  and (c)  $\theta_i = 50^\circ$ .

- Retrieval of leaf parameters**
  - ✓ Considered leaf DHR model : PROSPECT 5b (Féret et al, 2008)
  - ✓ In the 1000-2500 nm range:
    - $\tilde{p}_{PROSPECT} = \tilde{p}_{PROSPECT}(\theta_{dhr})$  avec  $\theta_{dhr} = (N, C_w, C_m)$
  - ✓ Combined PROSPECT+COSINE=PROCOSINE model:
    - $\tilde{R}_{hs, PROCOSINE} = \tilde{R}_{hs, PROCOSINE}(\theta)$  avec  $\theta = (N, C_w, C_m, b_{spec}, \theta_i)$
  - ✓ Model inversion based on iterative optimization of a merit function relating the measured and simulated spectra (3 parameters for PROSPECT, 5 parameters for PROCOSINE)

## Results

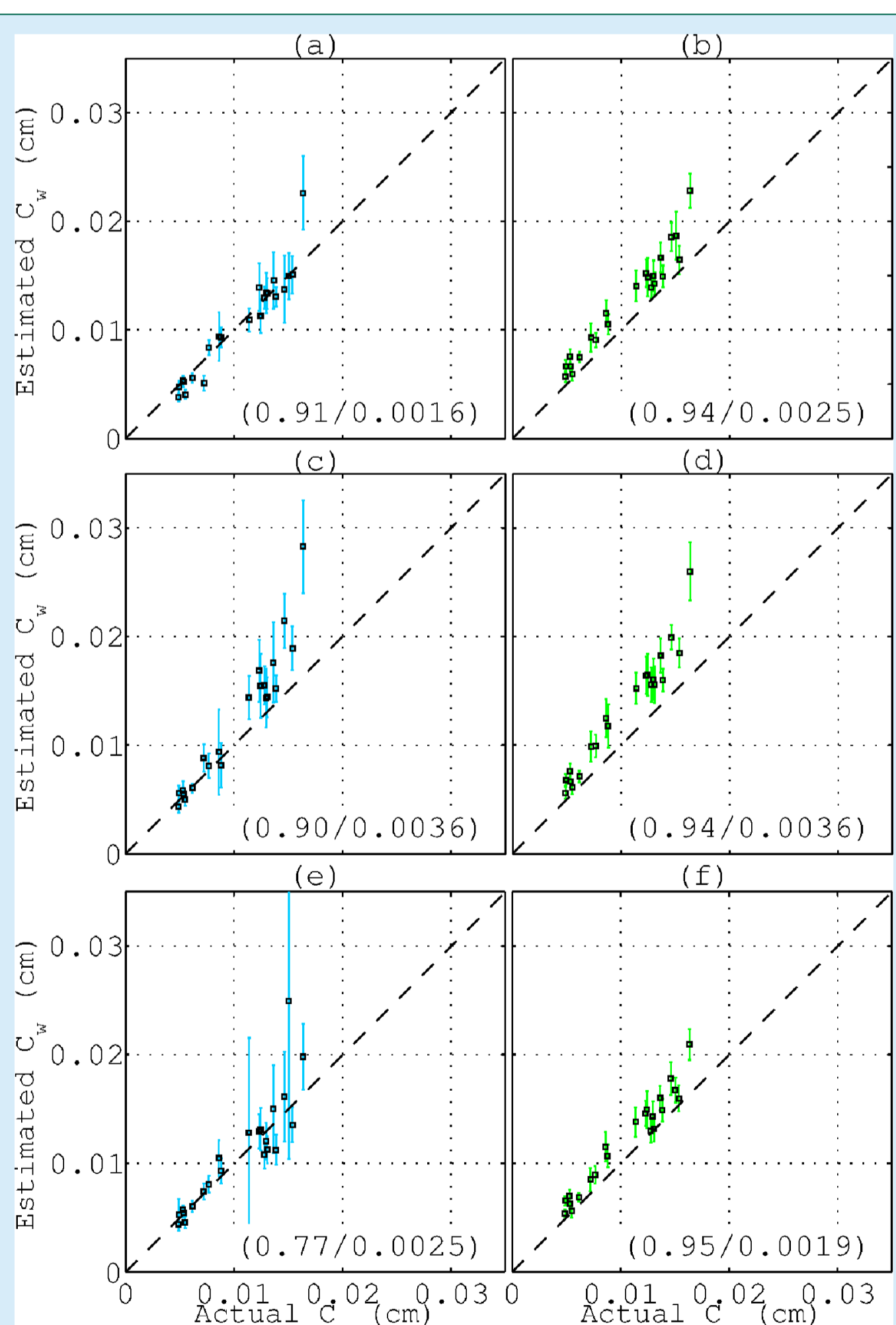


Figure:  $C_w$  estimation results (mean +/- std) obtained with PROSPECT (blue) and PROCOSINE (green) with  $\theta_i = 30^\circ$  (a-b),  $\theta_i = 10^\circ$  (c-d) and  $\theta_i = 50^\circ$  (e-f).  $R^2$  and RMSE are given in parentheses.

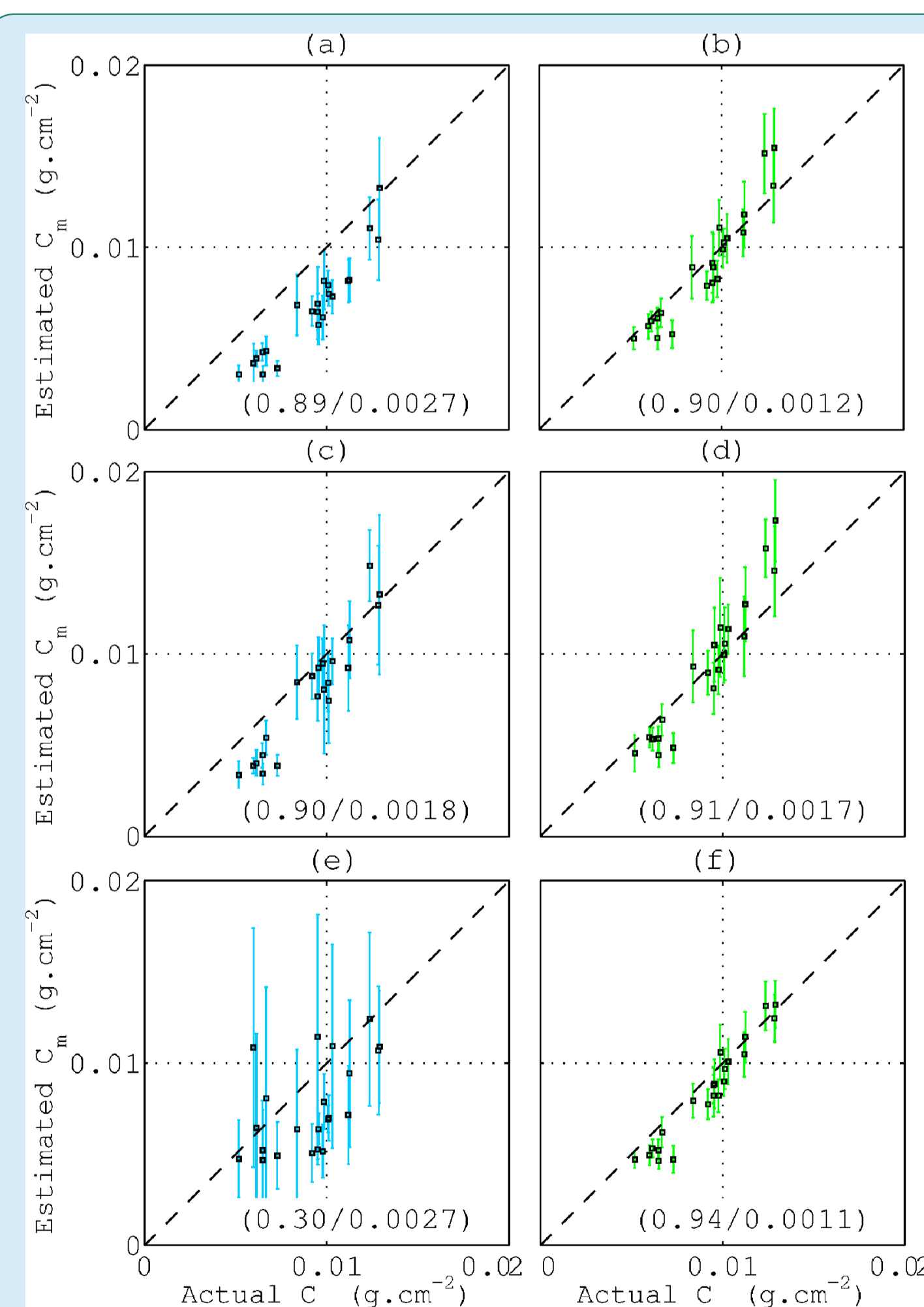


Figure:  $C_m$  estimation results (mean +/- std) obtained with PROSPECT (blue) and PROCOSINE (green) with  $\theta_i = 30^\circ$  (a-b),  $\theta_i = 10^\circ$  (c-d) and  $\theta_i = 50^\circ$  (e-f).  $R^2$  and RMSE are given in parentheses.

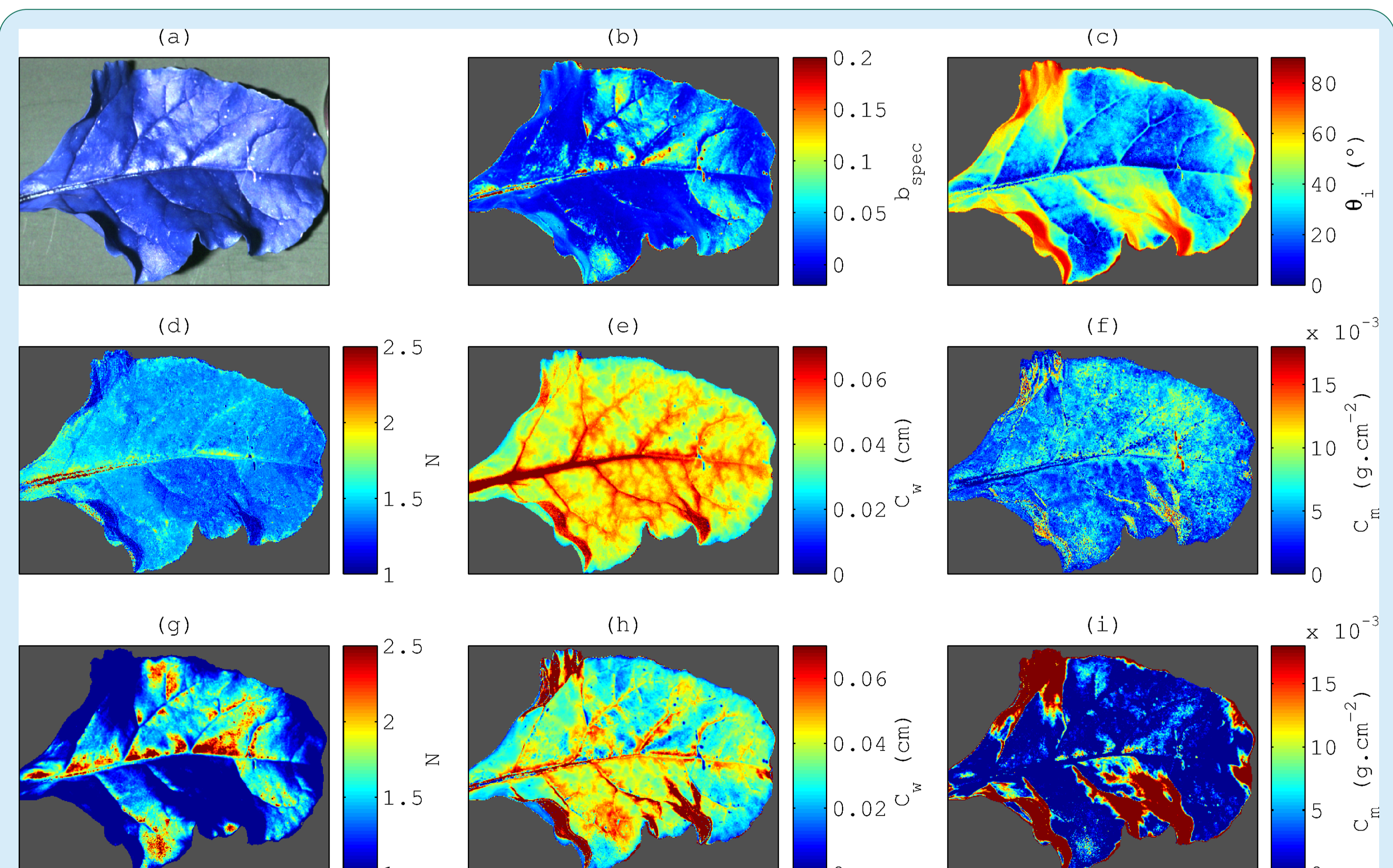
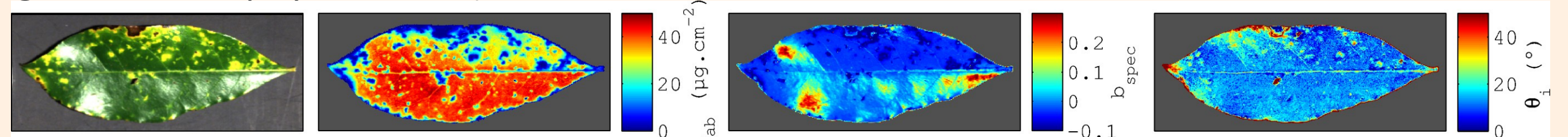


Figure: Estimated maps obtained from an extra image using PROCOSINE (b-f) and PROSPECT (g-i). This sugar beet leaf is characterized by a high  $C_w$  value and a low  $C_m$  value ( $C_w = 0.036$  cm,  $C_m = 0.004$  g.cm<sup>-2</sup>). (a) False color composite image (using bands 1458 nm, 2202 nm and 1662 nm), (b)  $b_{spec}$  (c)  $\theta_i$  (d) N, (e)  $C_w$  (f)  $C_m$  (g) N, (e)  $C_w$  and (f)  $C_m$ .

## Conclusions

- COSINE models specular reflection at the leaf surface and irregularities in leaf topography,**
- The combined PROCOSINE model allows the submillimetric retrieval of leaf parameters,**
- Also applicable to the VNIR range (400-1000 nm) for retrieving photosynthetic pigments (e.g., the chlorophyll content).**



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