

DEVELOPMENT OF SPECTRAL INDEX FOR DETECTION OF COFFEE LEAF MINER IN COFFEE

Research proposal

Humberto Fanelli Carvalho
Júlia Silva Morosini

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Humberto Fanelli **Carvalho**

Ph.D. candidate at IAC - Brazil



Júlia Silva **Morosini**

Ph.D. candidate at ESALQ/USP - Brazil

Introduction

- Coffee economic chain
 - 154 million bags exports – world wide
 - 160 US/lb – average future price - NY
- Coffee – gross revenue estimates 2017 - R\$19.56 bi
- Brazil – coffee leading exporter (estimates 44.83 million bags – 2017)



Introduction



Coffee leaf miner

- *Leucoptera coffeella* Guérin-Ménéville - lepdopteran
- 50% production reduction
- Principal control:



neonicotinoids



anthranilic diamide

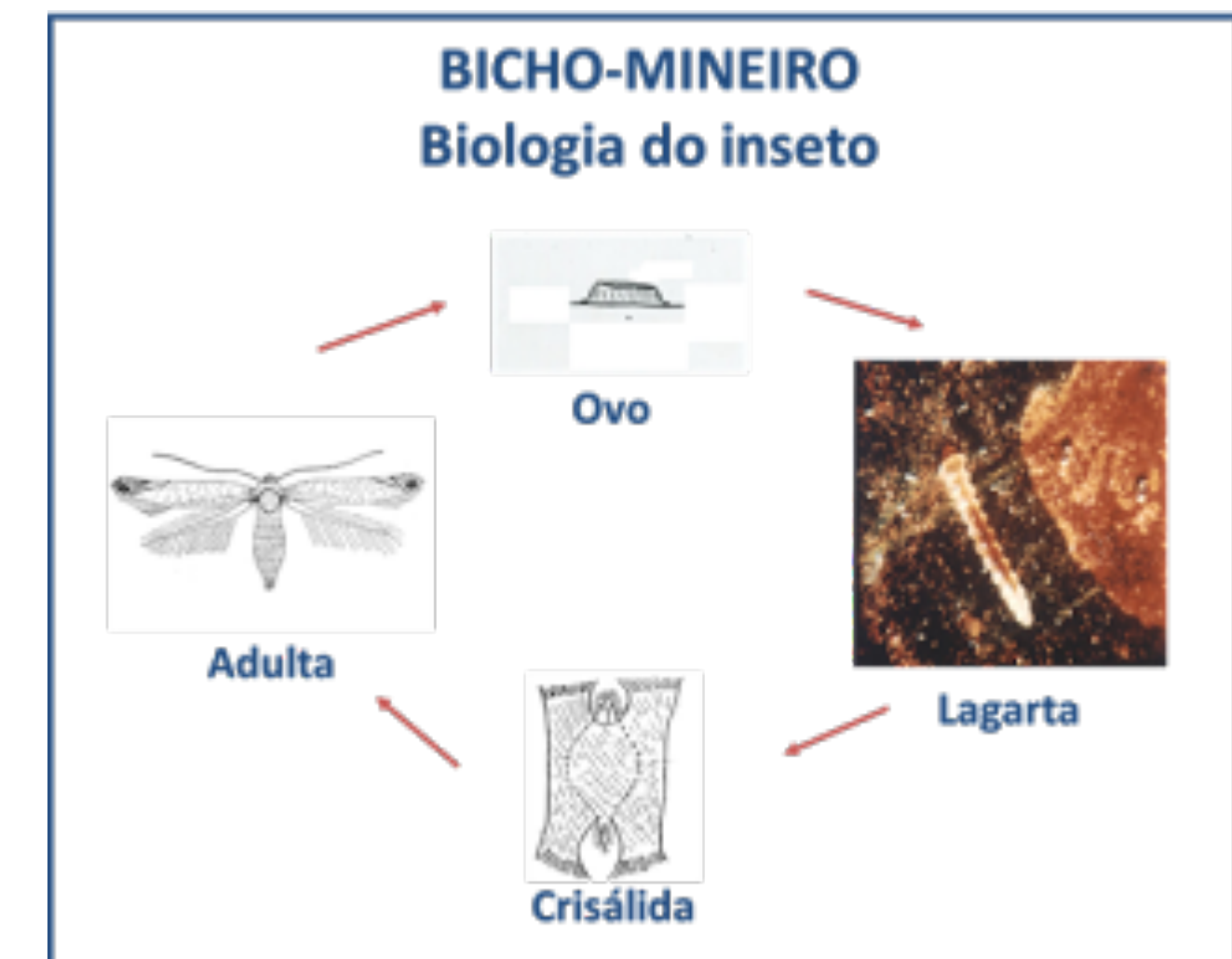


organophosphates

Introduction

Chemical control

- Environmental risk
- Risk to the operator
- Inefficiency of application in rough terrains
- Imbalance in the cycle pest population



Introduction

- Others: resistance cultivars (*Coffea racemosa*)
 - *Mean objective in breeding programs: pests and disease*



- Evaluation
 - Field
 - Green house



Coffee evaluation cycle

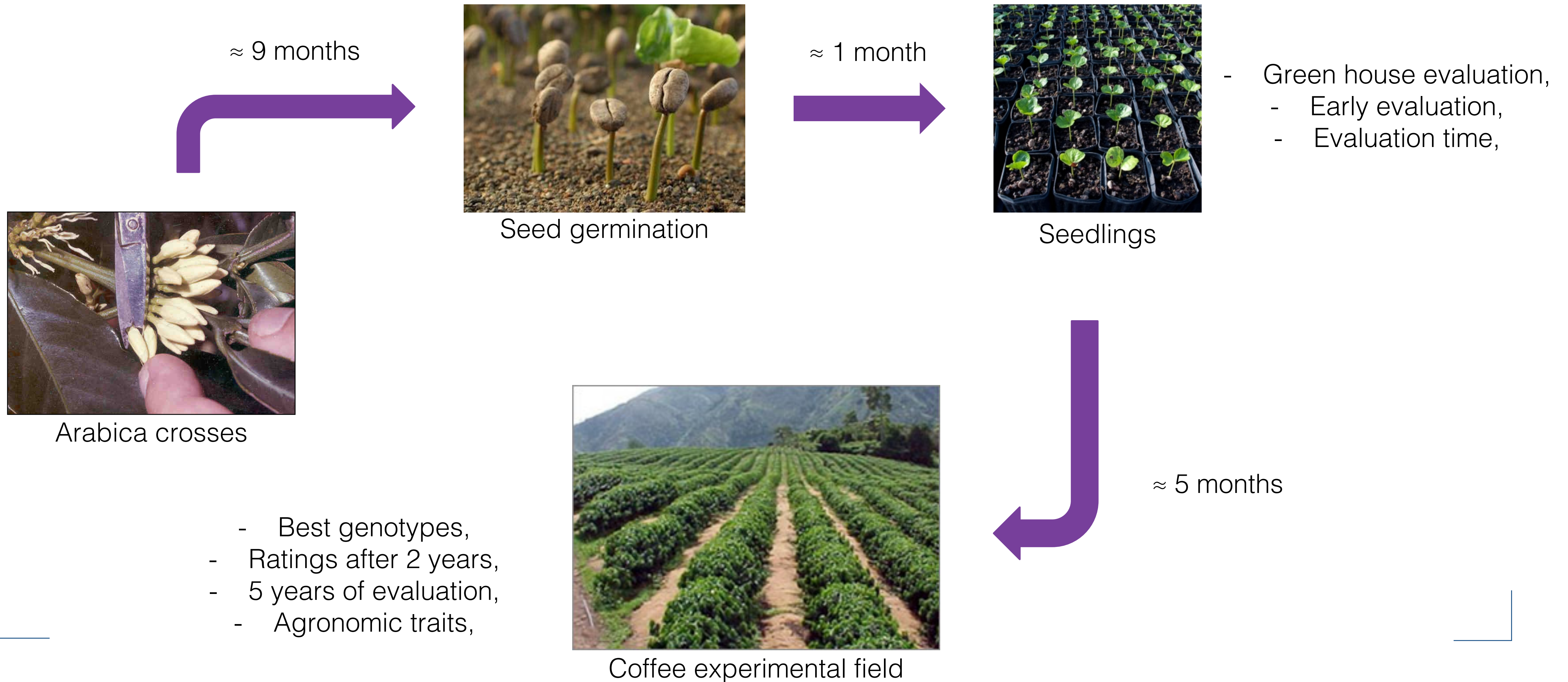
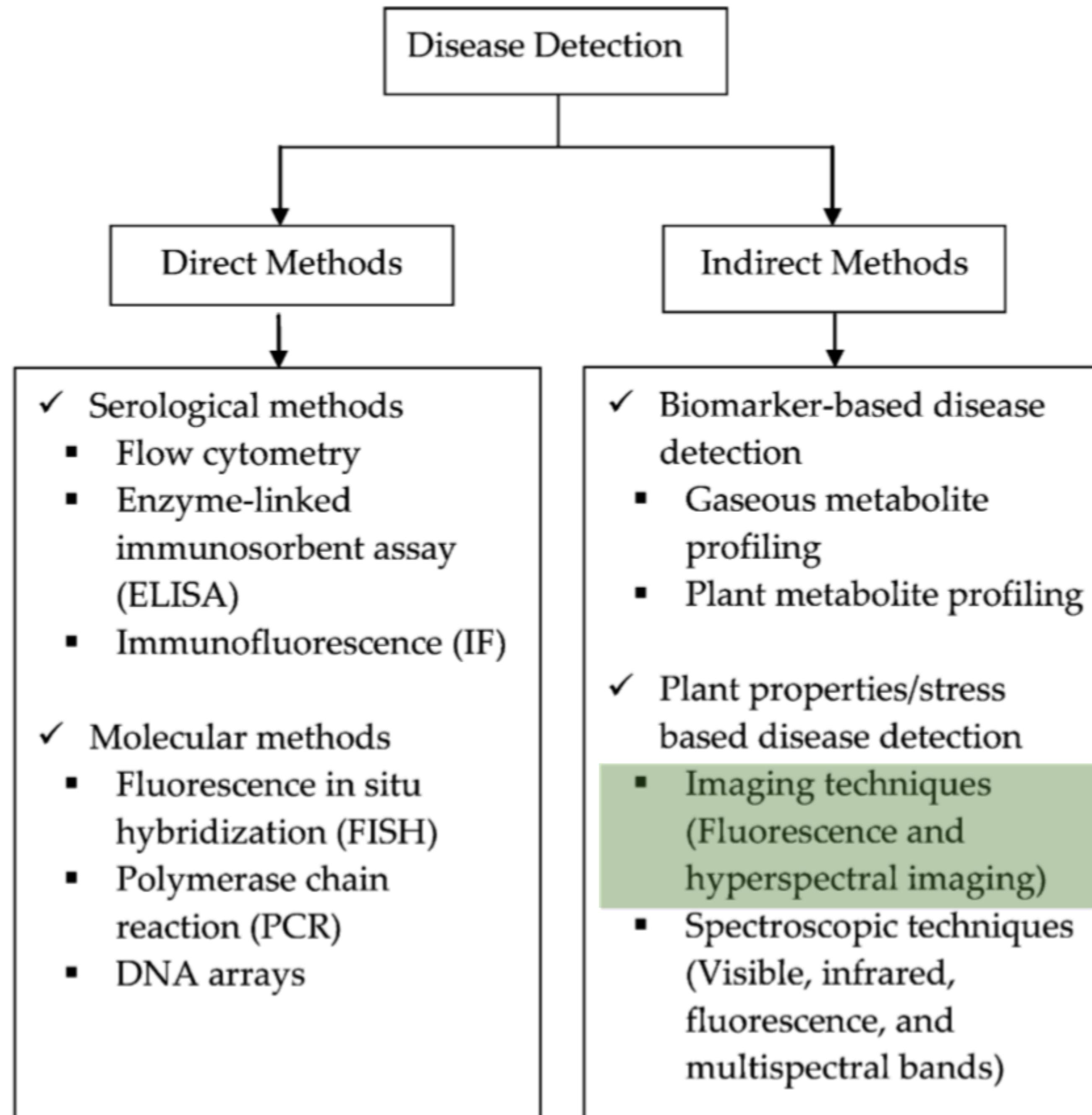


Image analysis

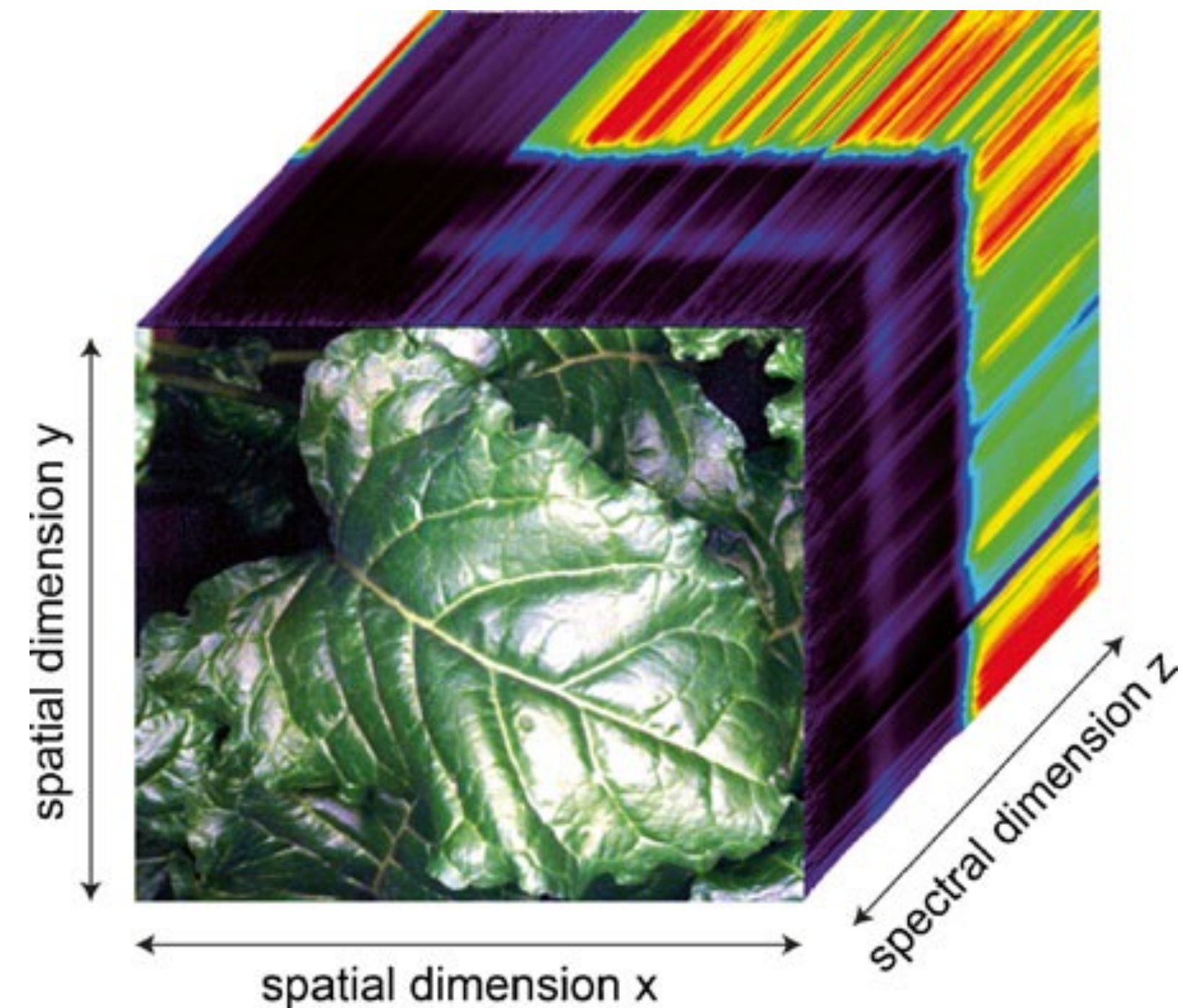
- Evaluations – note scale (Guerreiro-Filho, 1999):
 - Resistance
 - Severity of attack (notes for %)
- Note scale
 - Operational facilities
 - Subjectivity
 - Genotype categorization
- **Rate bias**



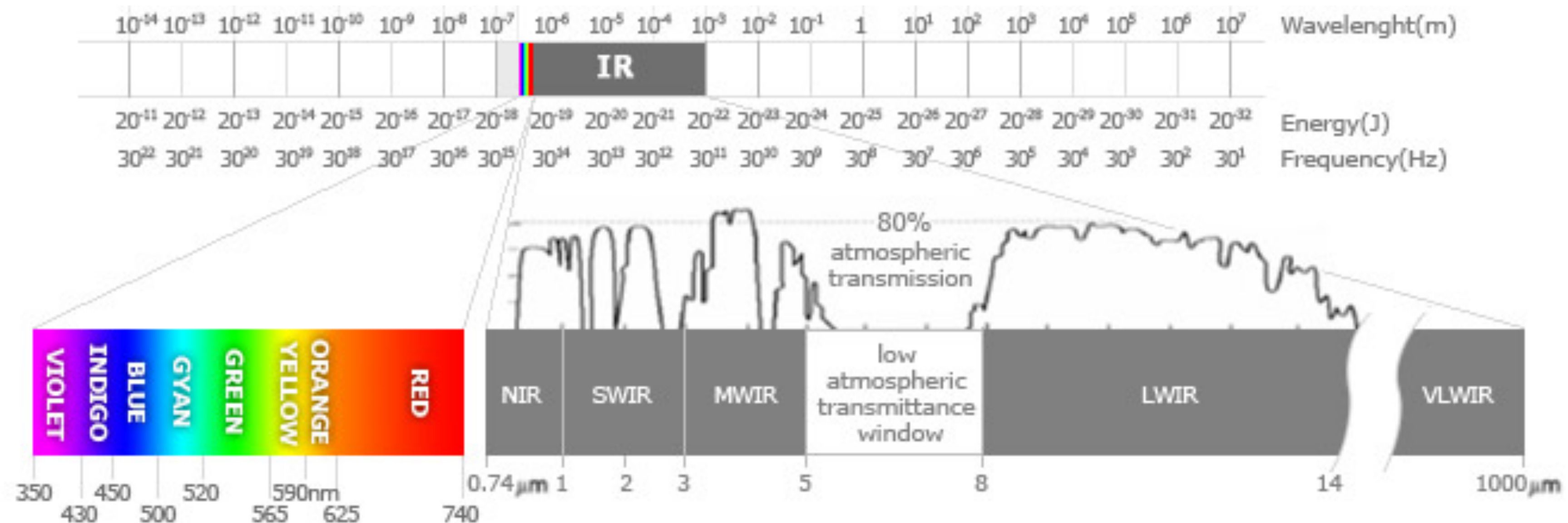


Argumentation

- **Hyperspectral cameras** facilitate the detection of both spectral and spatial information of an object:
- Each spatially located pixel of an image contains the full wavelength information



Argumentation



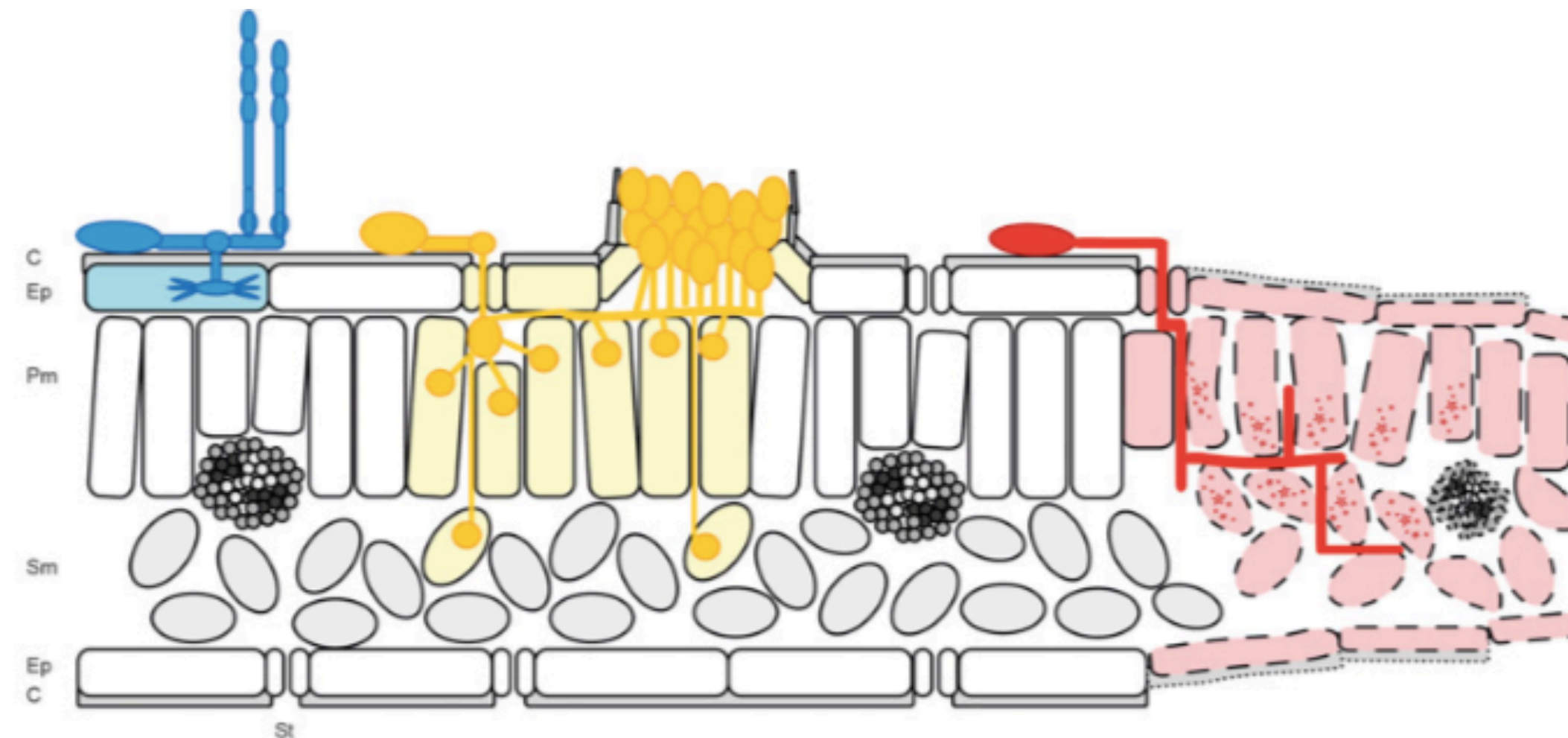
Leaf pigment content

- **Leaf structure**
- **Internal scattering processes**

- **Composition of leaf chemicals**
- **water**

Argumentation

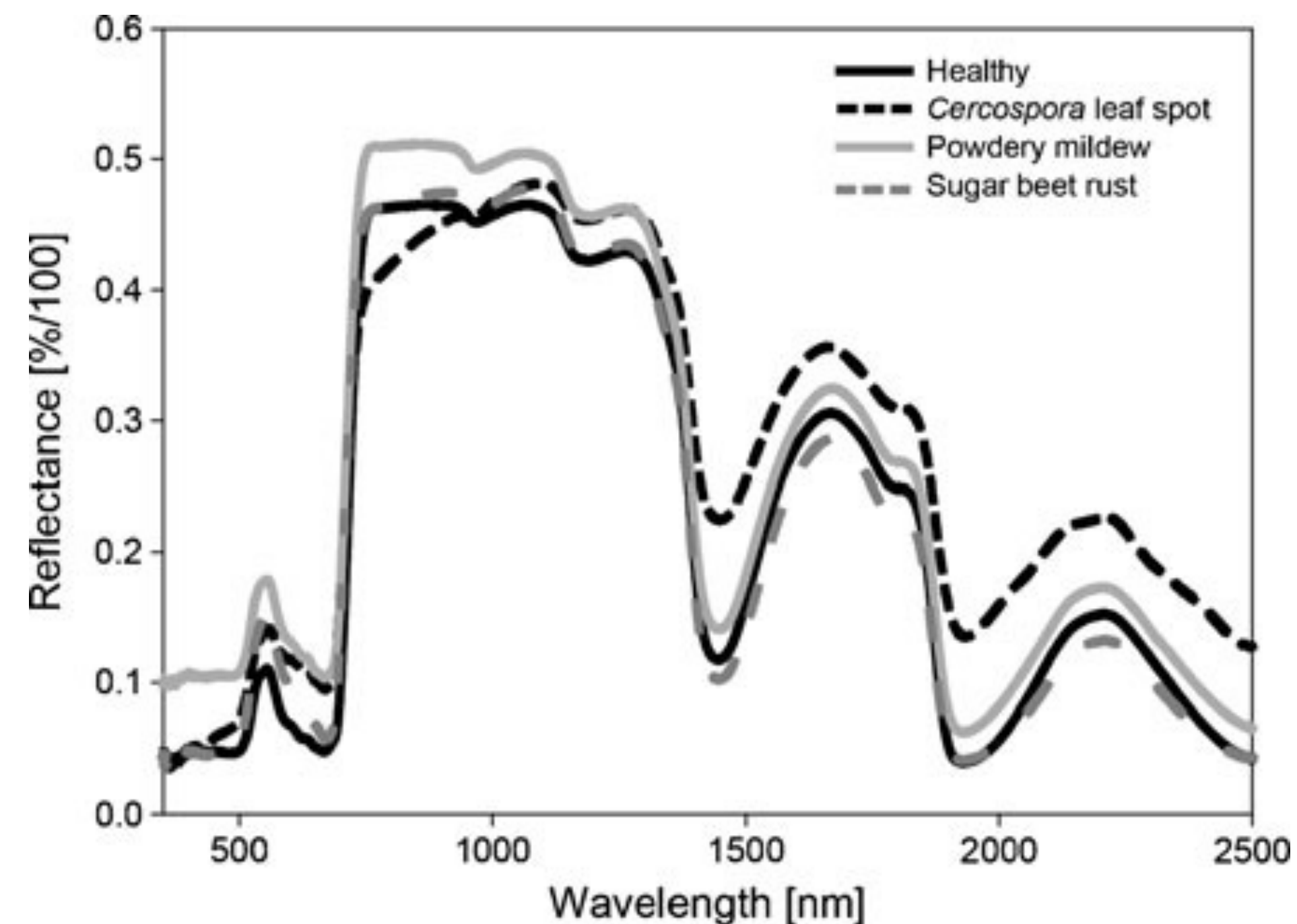
- Changes in reflectance: impairments in the **leaf structure** and **chemical composition** of the tissue that is highly specific.



- These highly complex and unique disease patterns enable an identification of diseases based on the spectral properties of plants.

Argumentation

- Construction of **spectral signatures**
- Disease-specific symptoms may be detectable



Argumentation

Analysis of sensor data

- Using specific wavelengths of spectral signatures, **spectral algorithms** have been developed to specify different plant parameters
- Benefits and disadvantages
- A combination of different wavelengths **spectral disease indices (SDI)** can be useful to simplify disease detection by spectral sensors

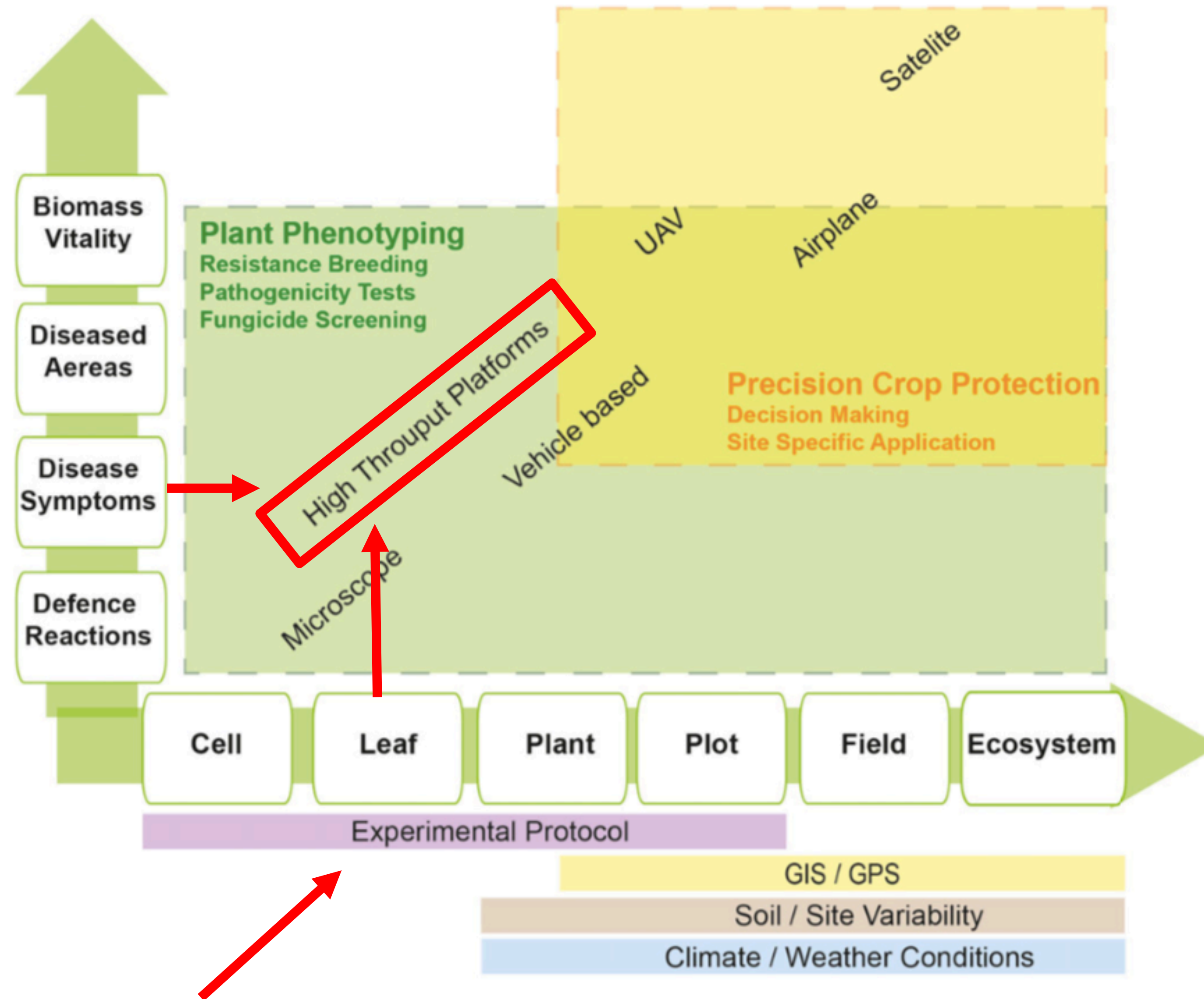


Motivation

- The identification of a disease, its discrimination from other diseases and abiotic stressors, using sensing techniques is still a challenge in vegetation monitoring;
- Interpretation of spectral reflectance data without knowledge on spectral characteristics of leaves and typical symptoms is impossible at present

Development of spectral index for detection of coffee leaf miner in coffee

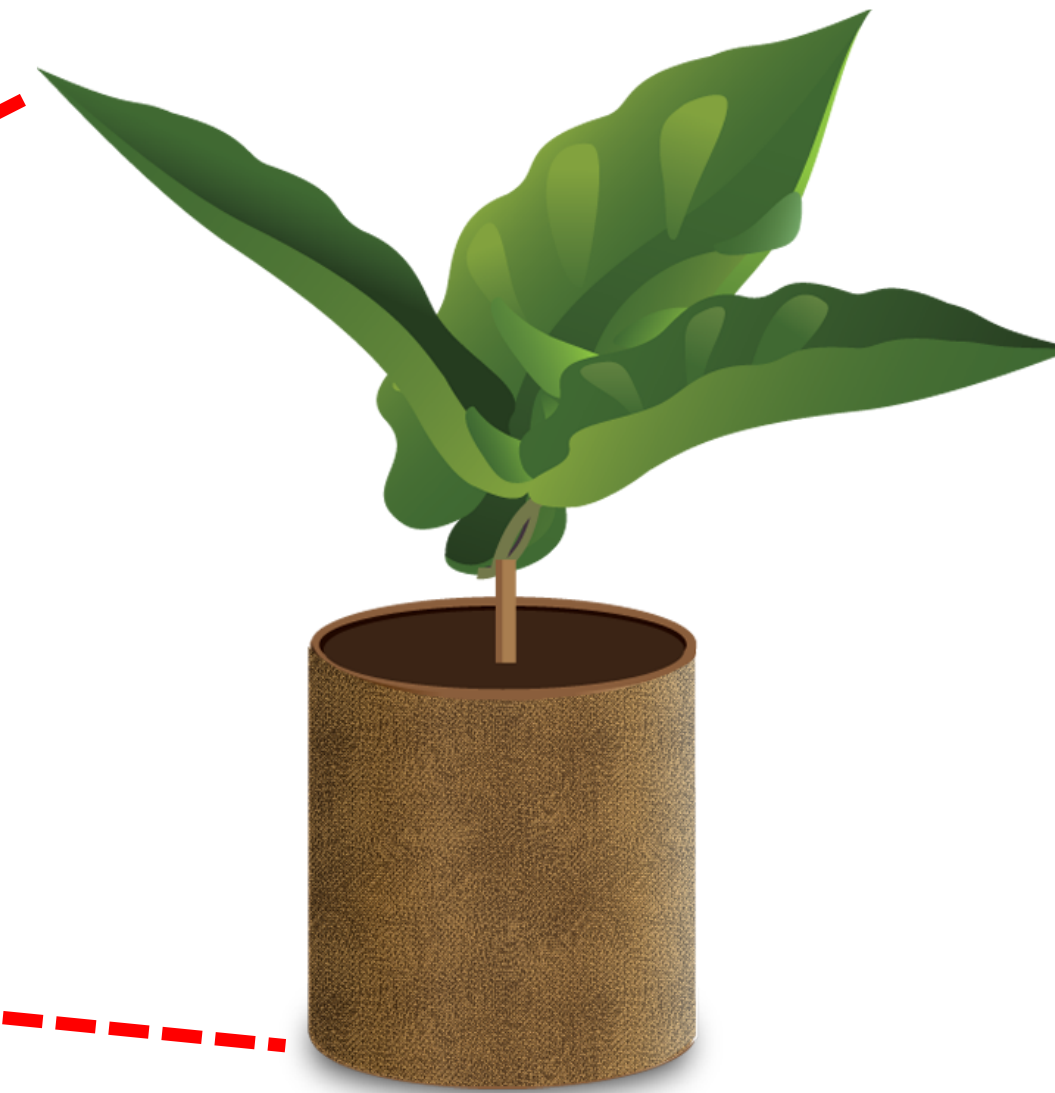
Methodology



Methodology

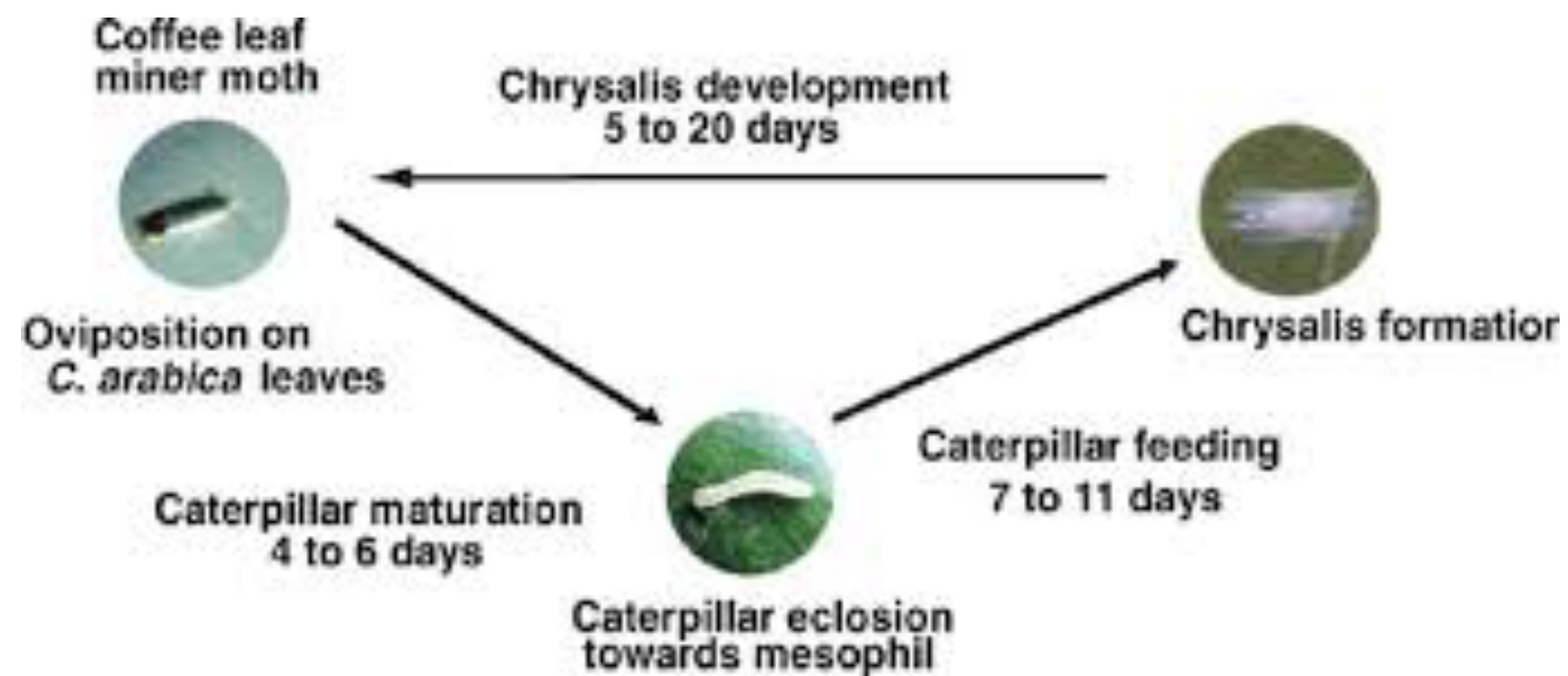
Experimental setup

- Use a data set of spectral signatures from healthy coffee leaves and from coffee leaves infested with LM



Methodology

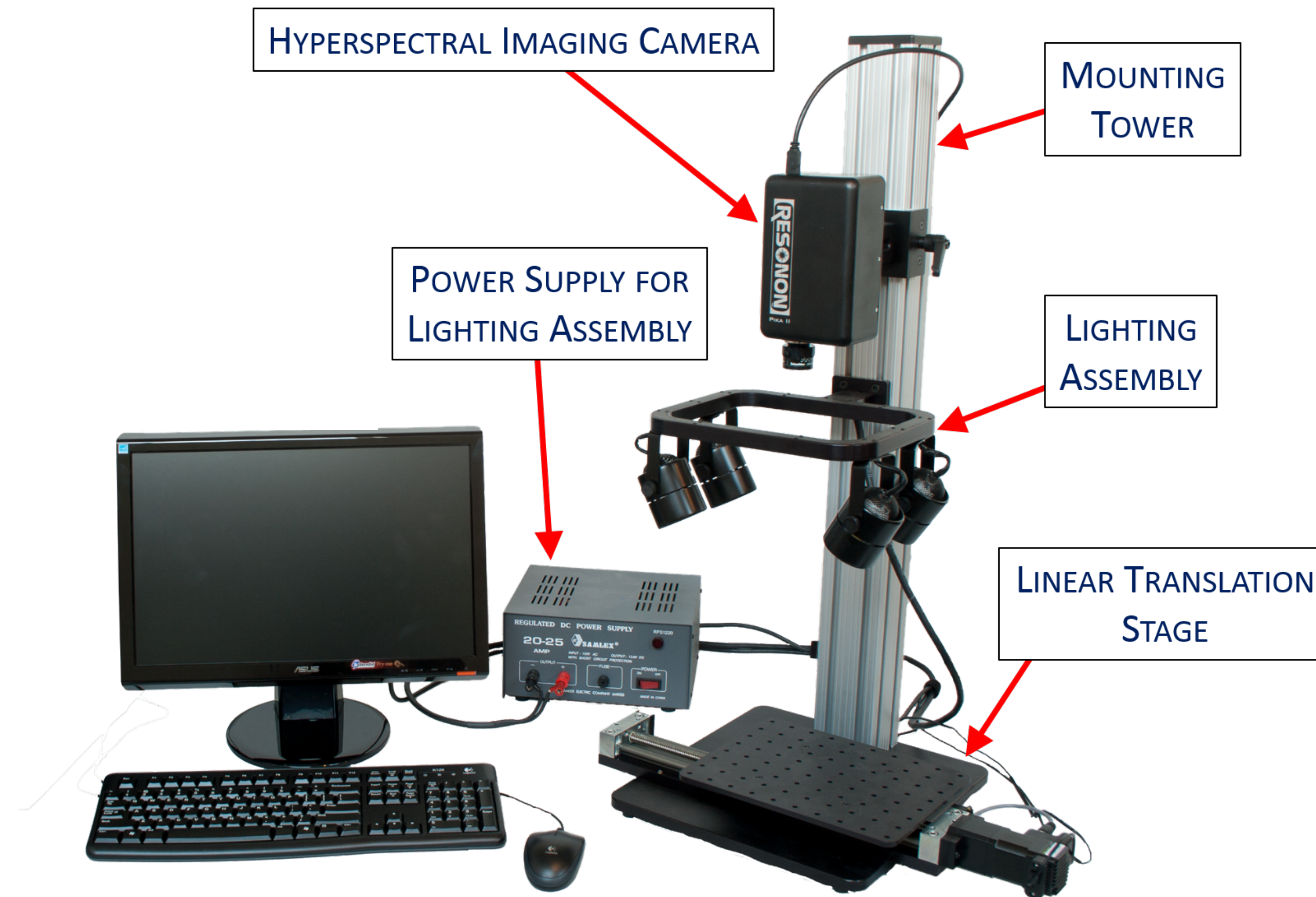
Experimental setup



- Controlled conditions
 - Isolated stressor
 - Lightening and background pattern
- Two experiments

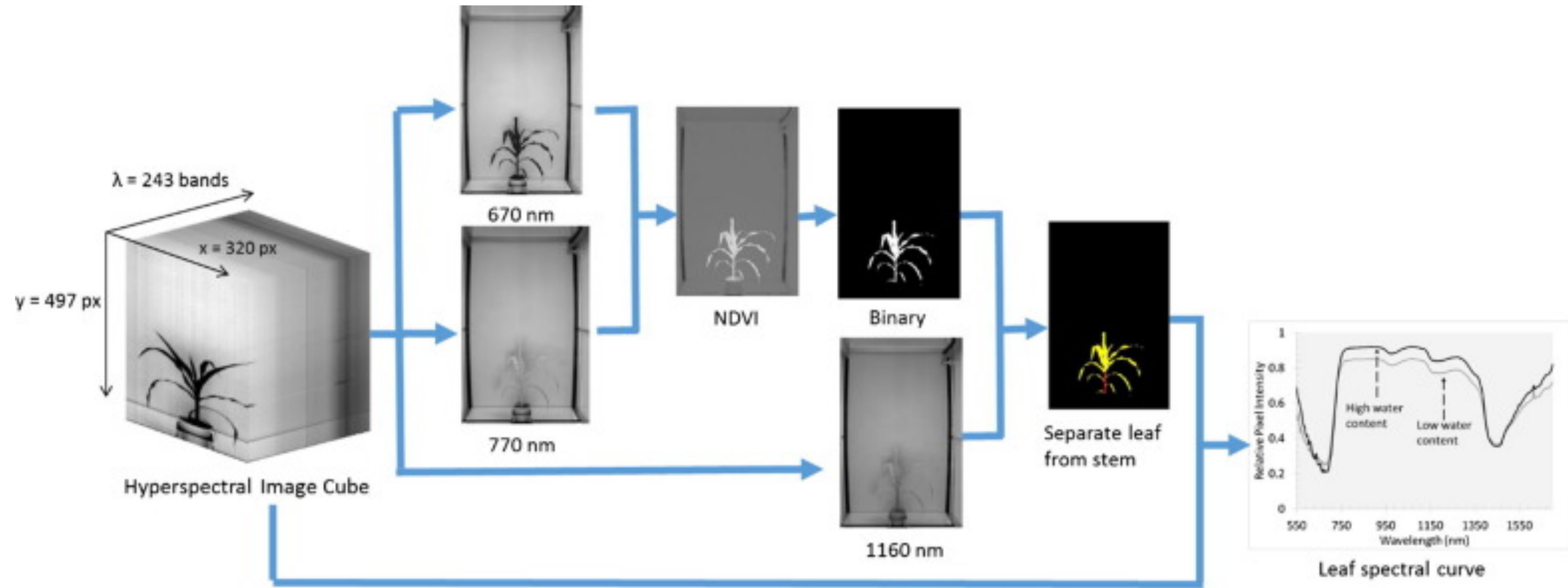
Methodology

Measurement of leaf reflectance



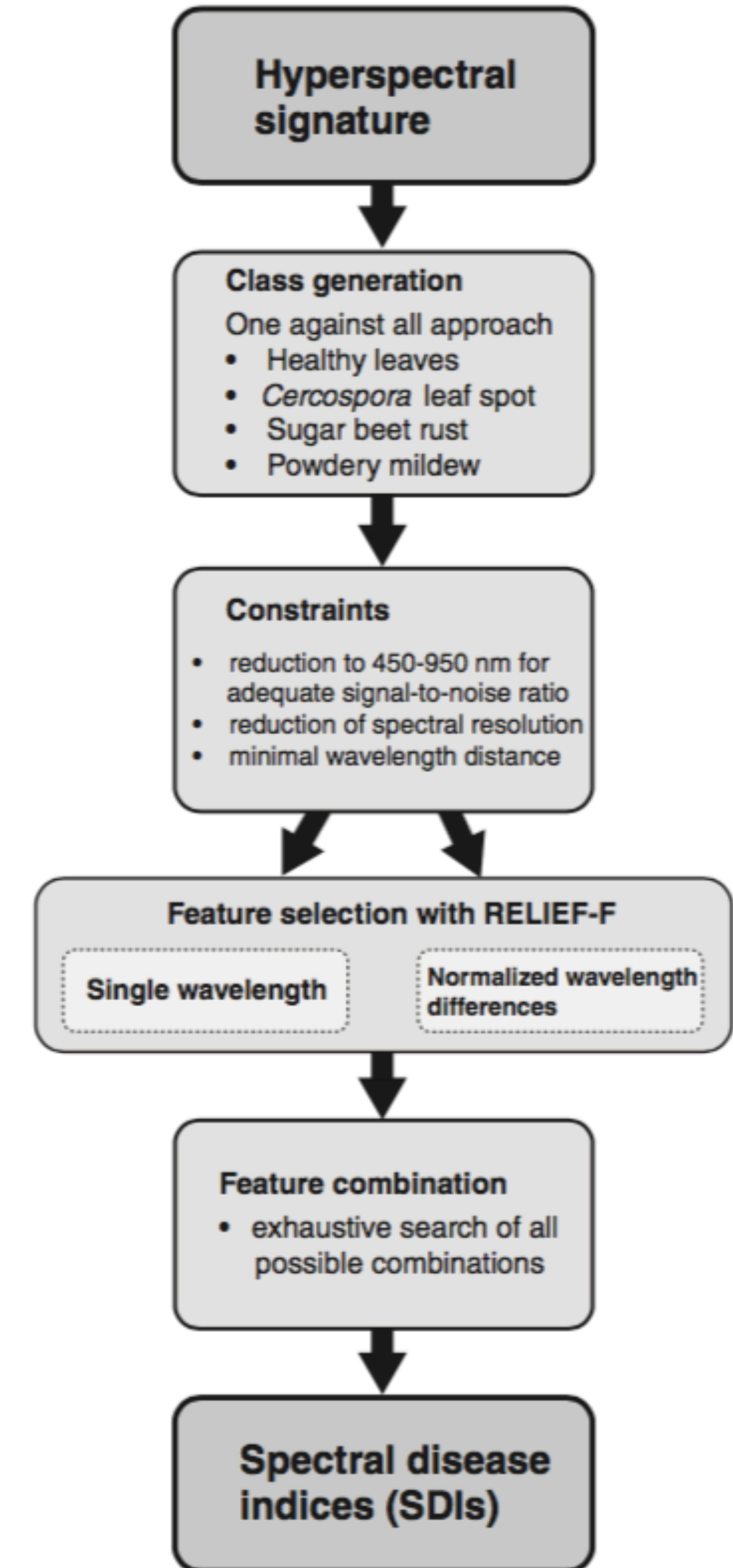
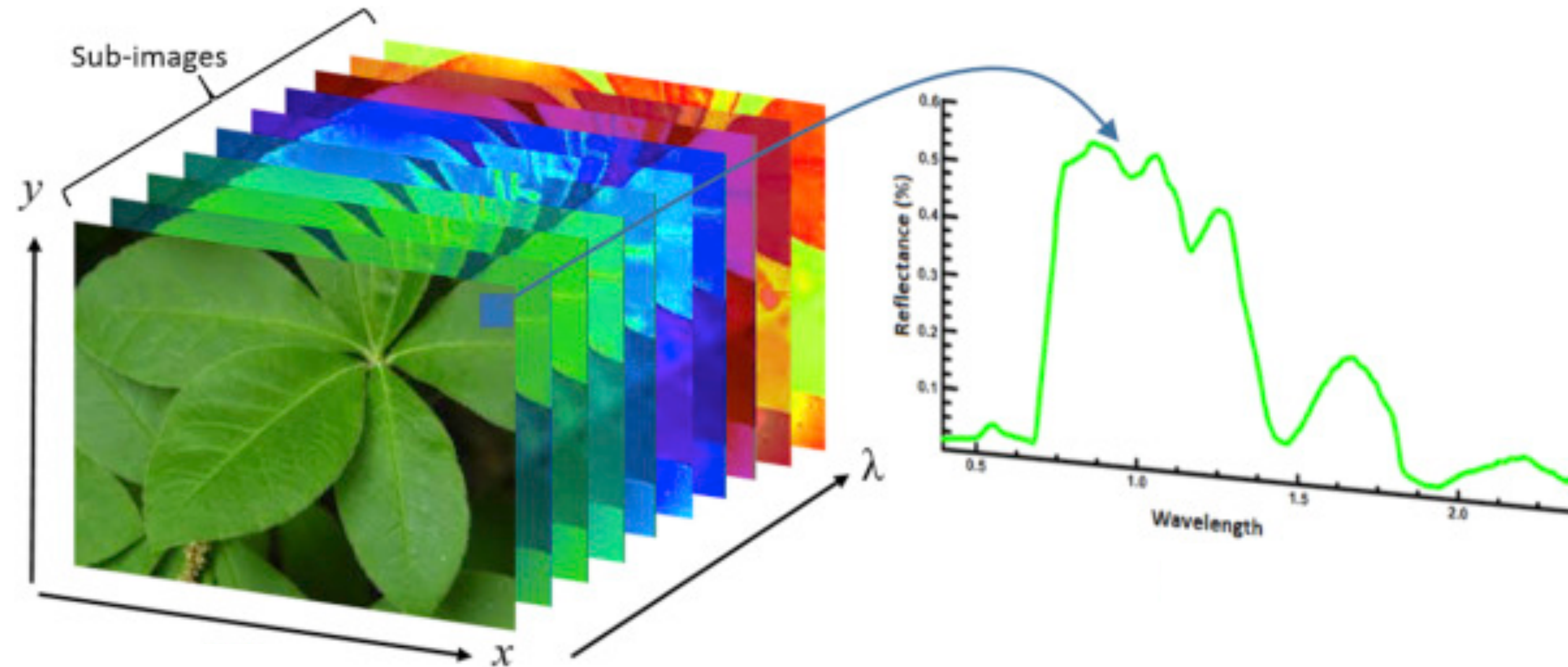
Methodology

Measurement of leaf reflectance



Methodology

Construction of SDI



Methodology

Construction of SDI

- RELIEF-F algorithm: most relevant single wavelengths and wavelength differences
 - estimates the relevance of wavelength according to their goodness to separate samples of both classes which are near to each other.

$$\text{Healthy} - \text{index (HI)} : \frac{R534 - R698}{R534 + R698} - \frac{1}{2} \cdot R704$$

$$\text{Cercospora leaf spot} - \text{index (CLS)} : \frac{R698 - R570}{R698 + R570} - R734$$

$$\text{Sugar beet rust} - \text{index (SBRI)} : \frac{R570 - R513}{R570 + R513} + \frac{1}{2} \cdot R704$$

$$\text{Powdery mildew} - \text{index (PMI)} : \frac{R520 - R584}{R520 + R584} + R724$$

Methodology

Discriminant analysis of LM symptoms level



0 - 30%



30 - 60%

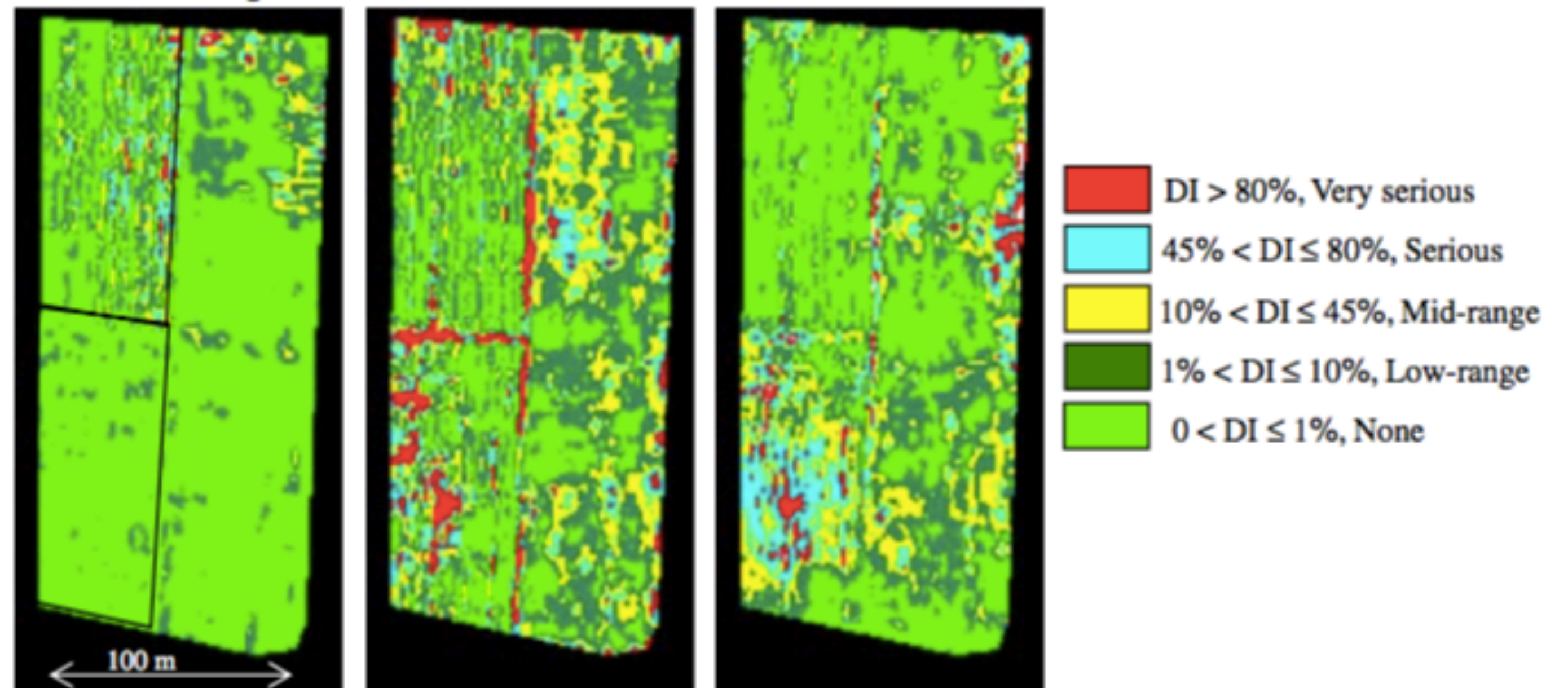


60 - 100%

Methodology

Discriminant analysis of LM level

- Supervisioned ML
 - SVM (Singh et al. 2016)
 - PLS-DA (Chemura et al. 2017)
- K-means (Kaur et al. 2016)



Estimated costs

Features	Price (R\$)
Experiment installation	300.00
Camera acquisition	17,000.00
Imagery structure	200.00
TOTAL	17,500.00



SOC710

Commercial name	Chemical class	Dosage	Average Price	Price/ha
Actara (250WG)	Neonicotinoids	1kg/ ha	R\$ 293/kg	R\$293,00
Altacor	Anthranilic diamide	90g/ha	R\$ 840/pct 450g	R\$168,00
Sabre	Organophosphates	2L/ha	R\$ 176/5L	R\$70,4

Expected results

Current LM monitoring

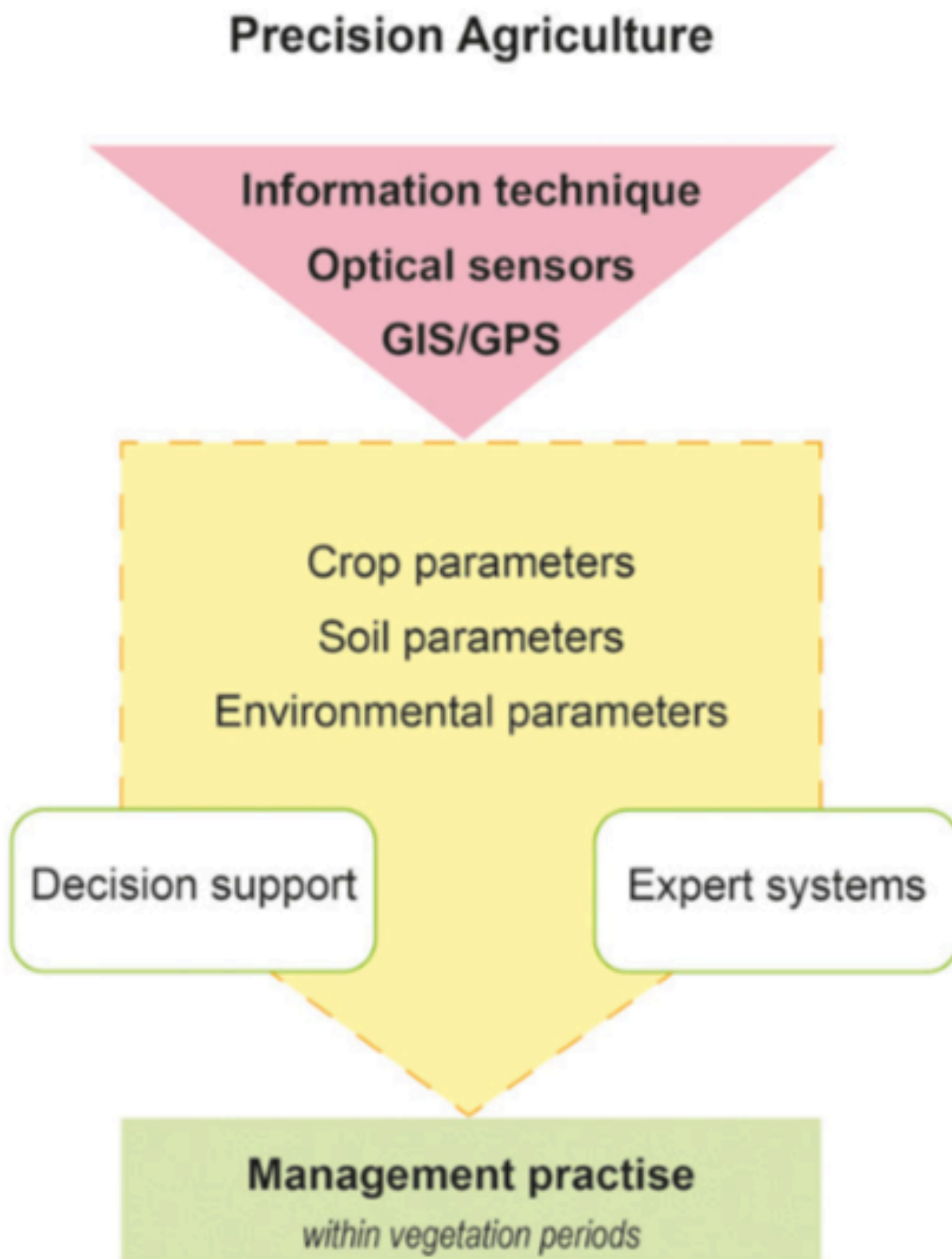
- Occasional field surveys
 - teams of specially trained and experienced personnel;
 - strenuous especially for large coffee plantations;
 - subjective.

Proposed LM monitoring

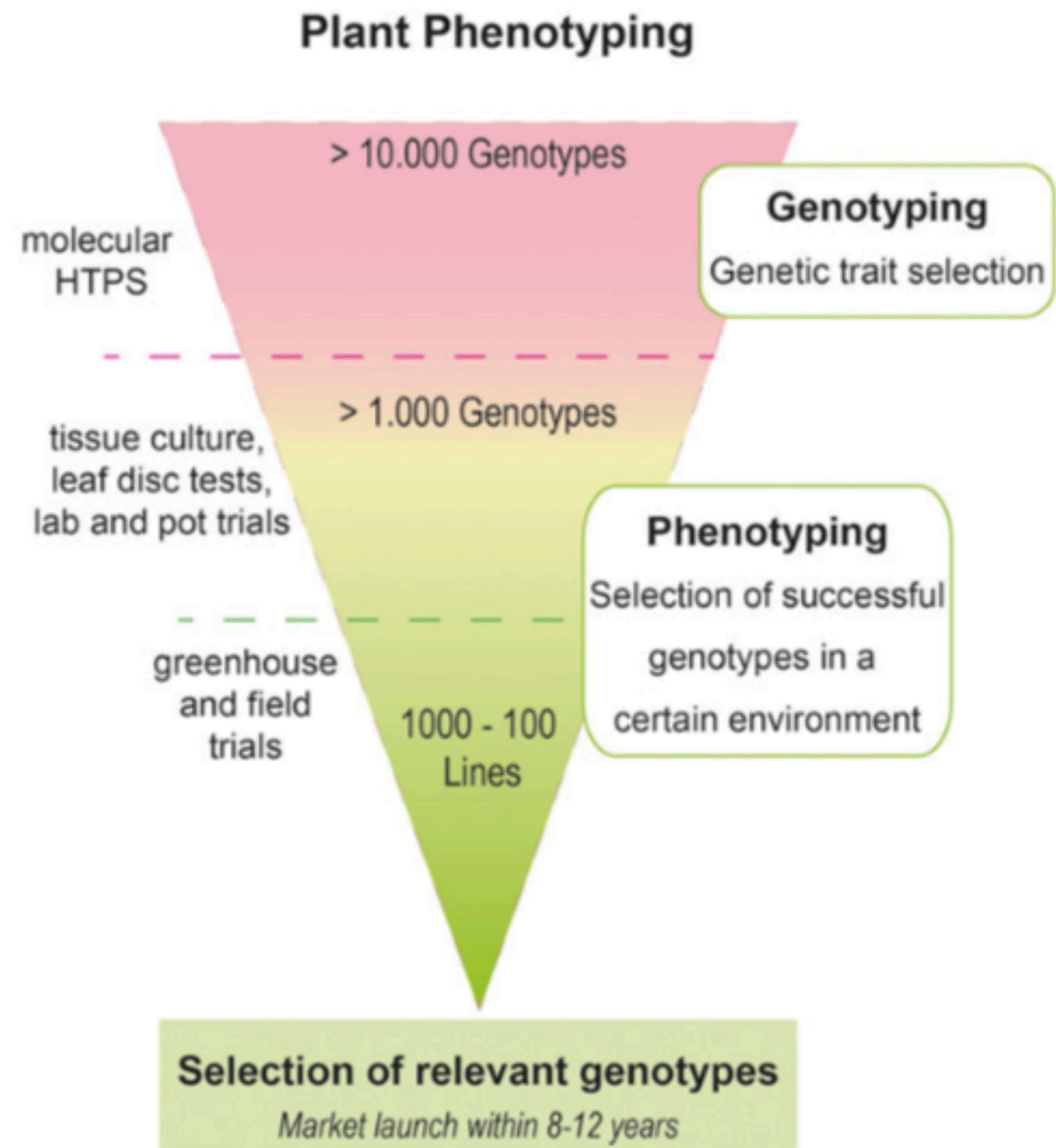
- Proximal/remote sensing
 - timely and spatially assessment of plant condition;
 - throughout the growing season;
 - non-invasive;
 - provision of accurate information.

Expected Results

- Hyperspectral techniques, alone and in combinations with sensor fusion approaches can highlight host-pathogen interactions at an **early stage**
- Management of the complexity of an identification and quantification of different kinds of stressors
- Decrease of yield loss by LM
- Sustainable impact
- **Potential to be extended to detecting and monitoring under field conditions**
- **Detection of resistant genotypes in due course (plus genomic data)**



vs.





THANK YOU!