

Toward Machine Learning Implementation in AI for Agro- Food Sustainability From Olive Oil Quality to Citrus Fruit Detection in the field

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Politecnico
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DI PALERMO



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Confederazione Svizzera
Confederaziun svizra





Workshop at Alahambra Beer in Granada



Measurement Campaign at Granada



SSP lectures at UGR



Secondment at TOELT LLC (CH)

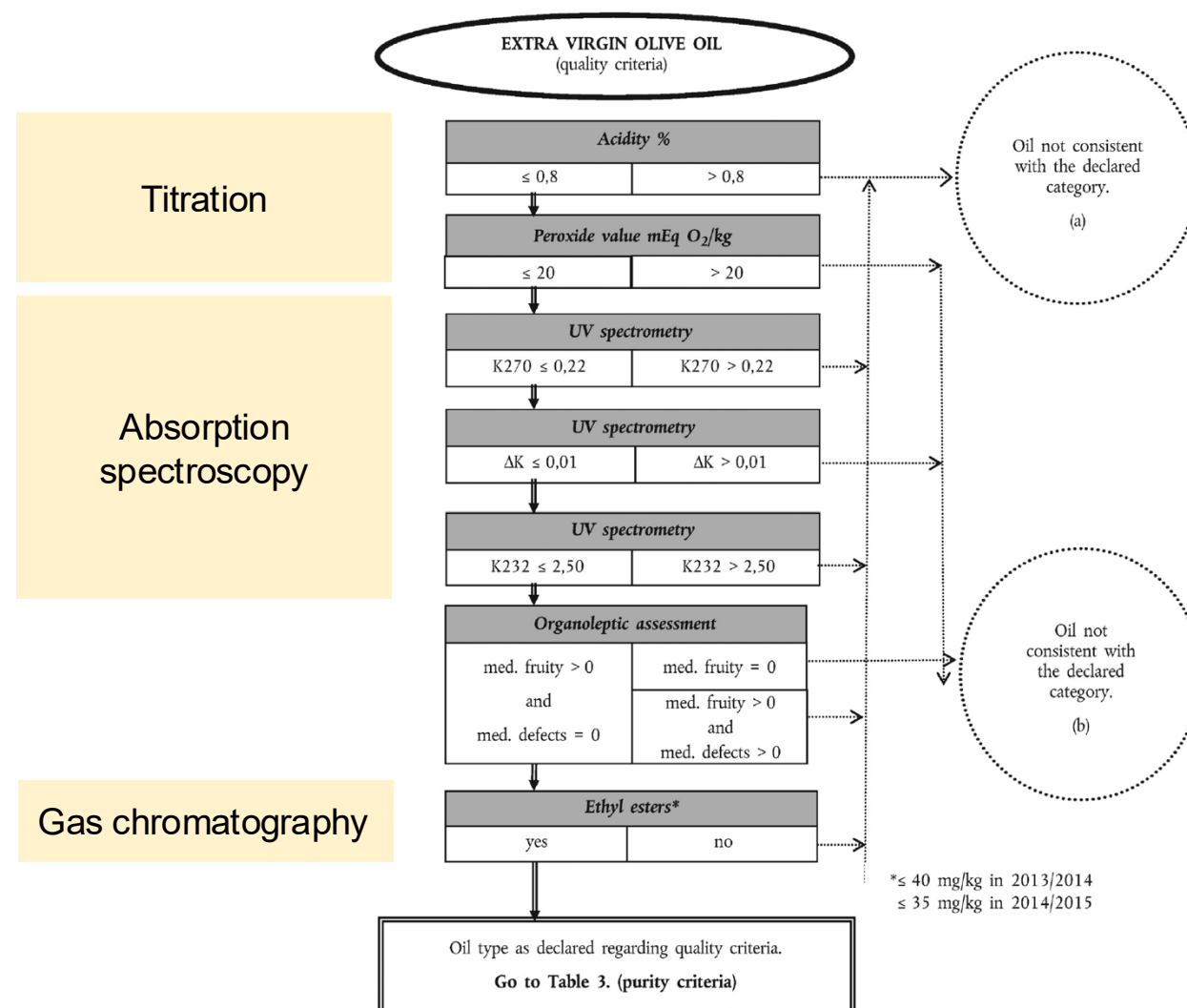


Secondments in Sicily (Palermo)

EVOO quality



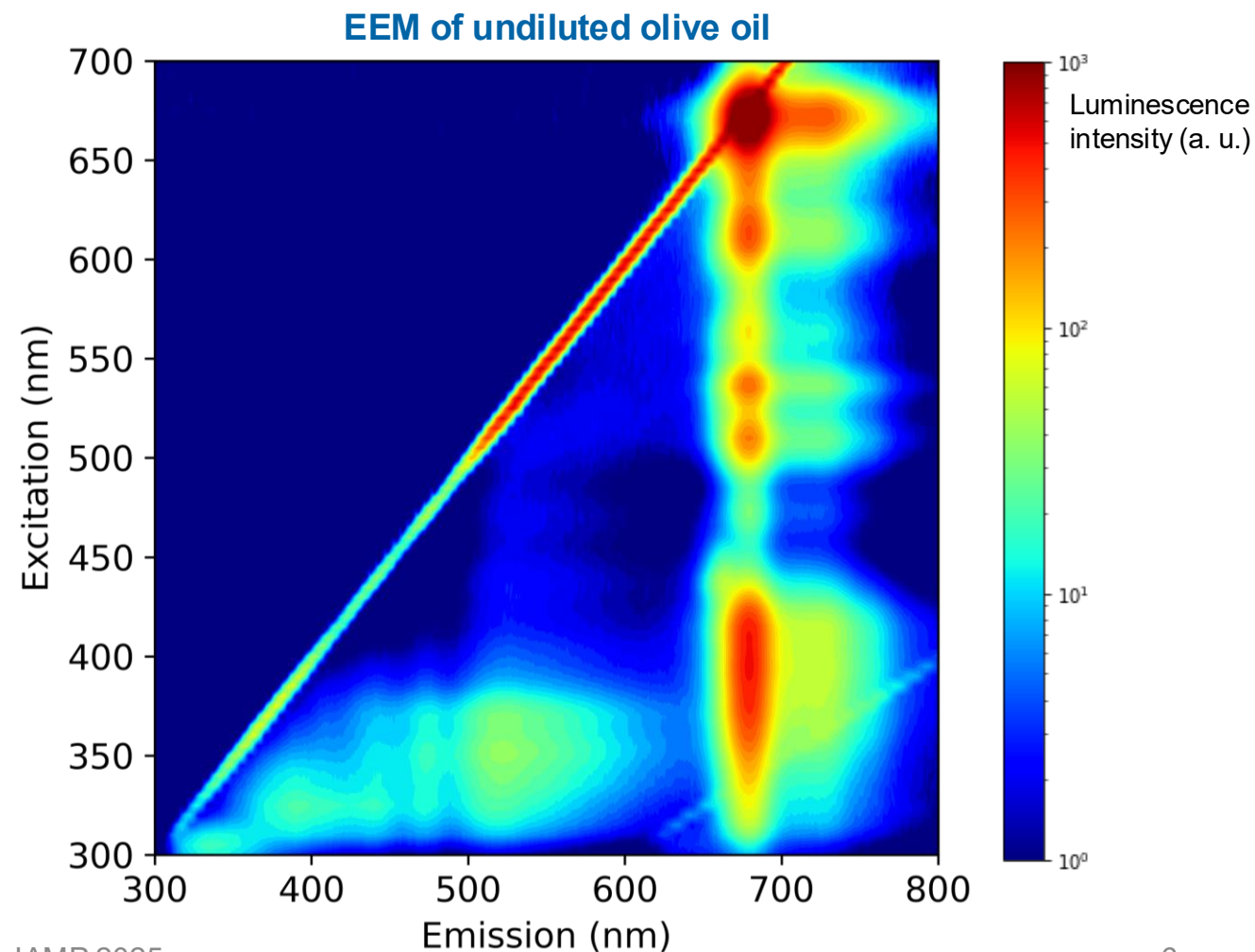
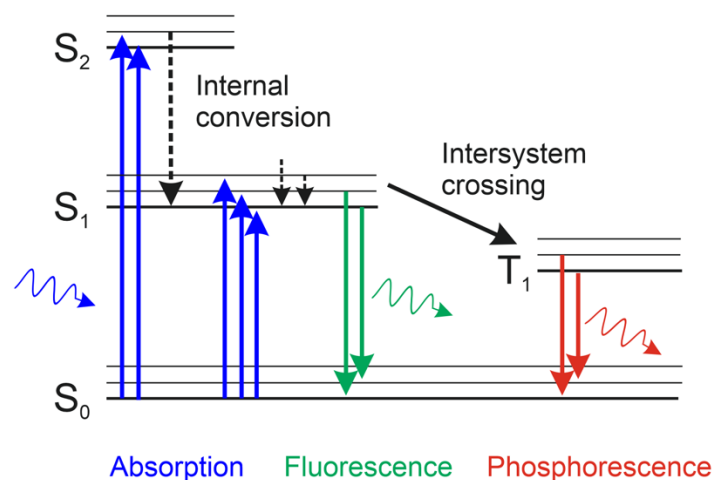
Quality assessment of EVOO



Why fluorescence?

Fluorophores naturally present in oil:

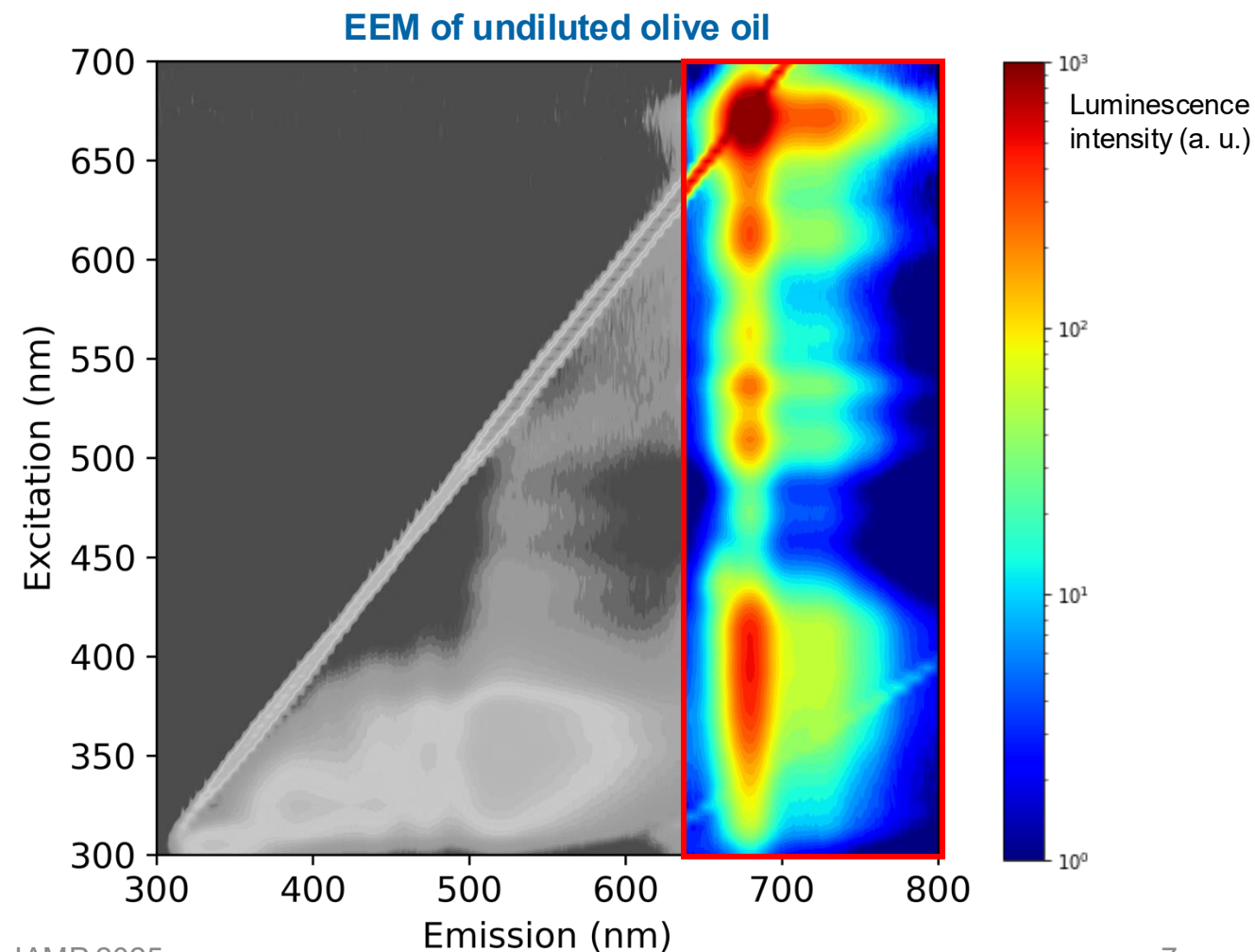
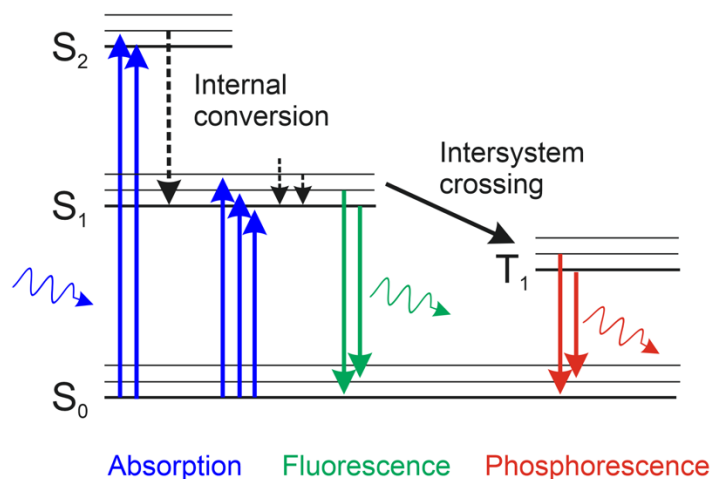
- Chlorophylls *a*, *b*, *d*
- Flavonoids, carotenoids
- Polyphenols
- Tocopherol, vitamin E
- Oxidation products



Method: fluorescence

Fluorophores naturally present in oil:

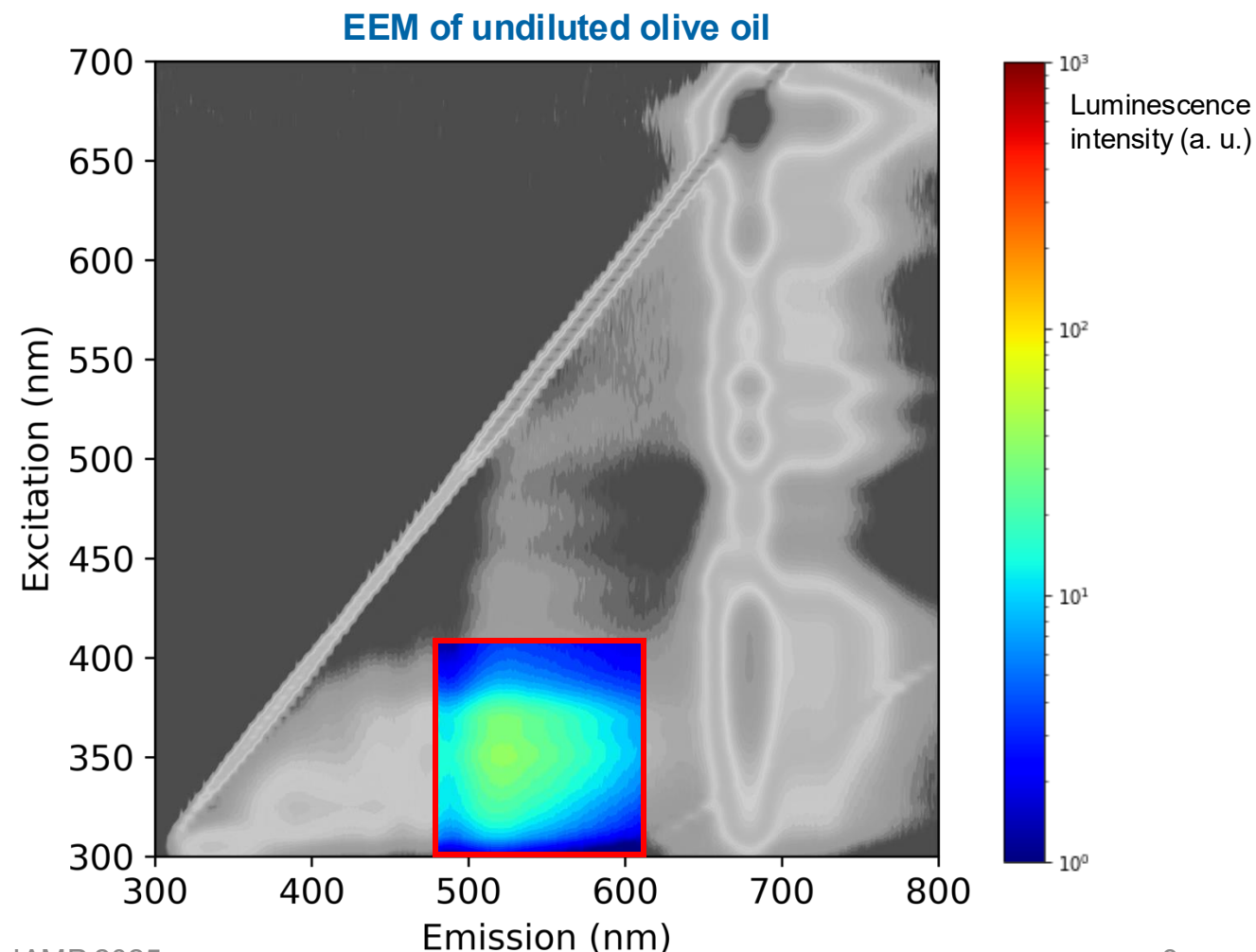
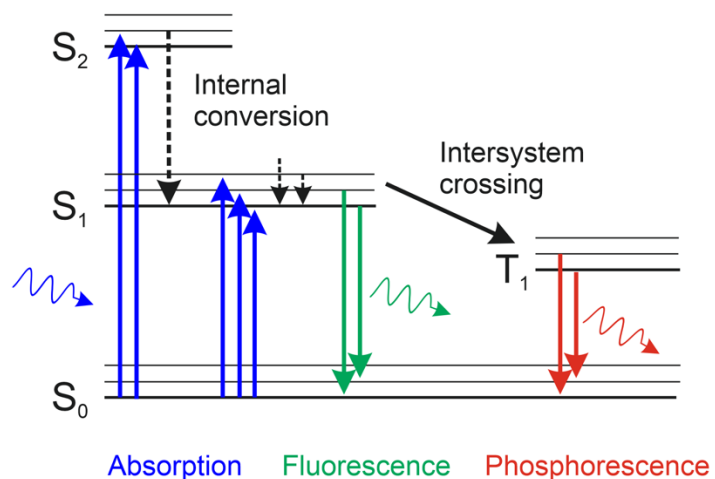
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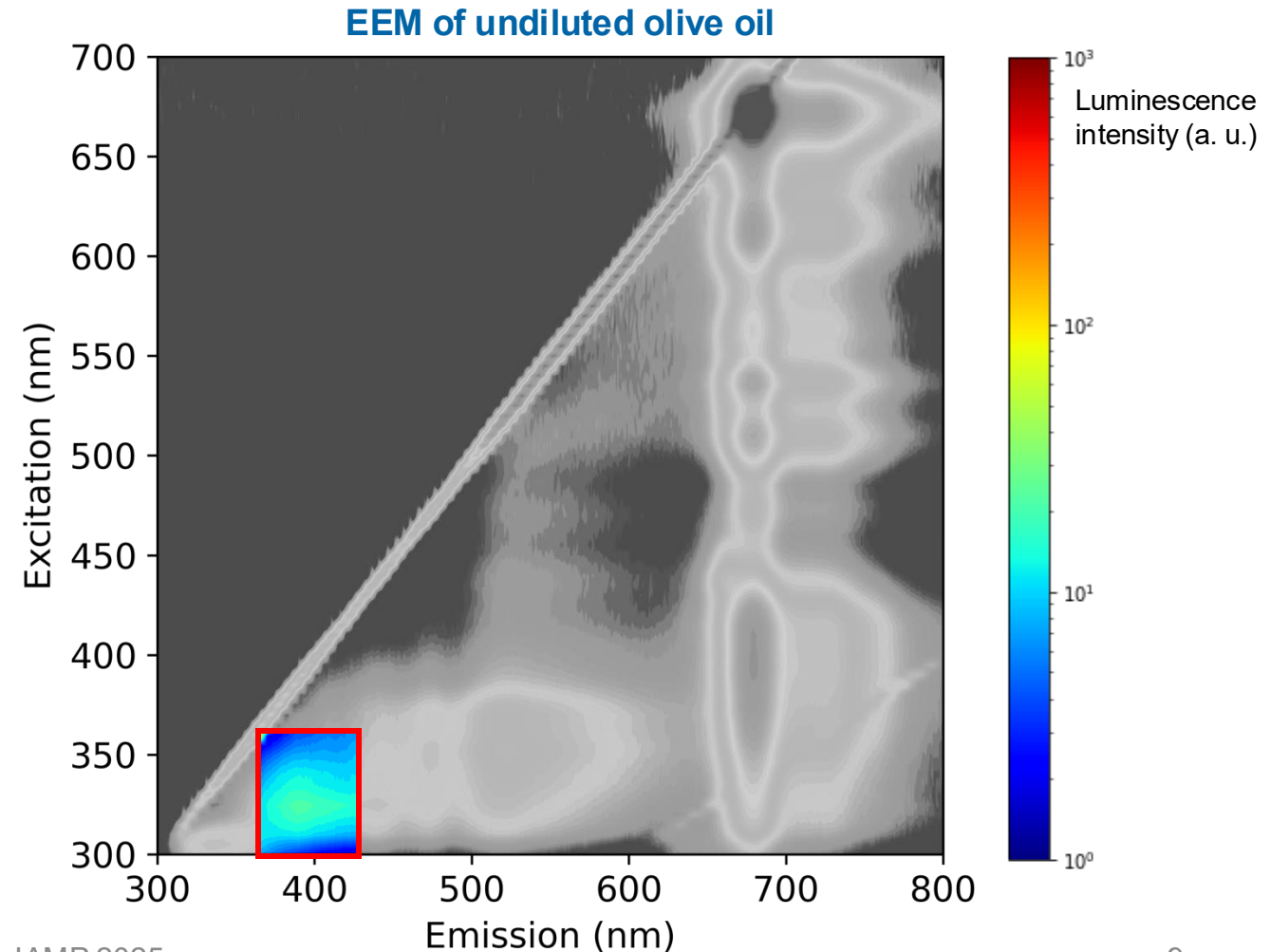
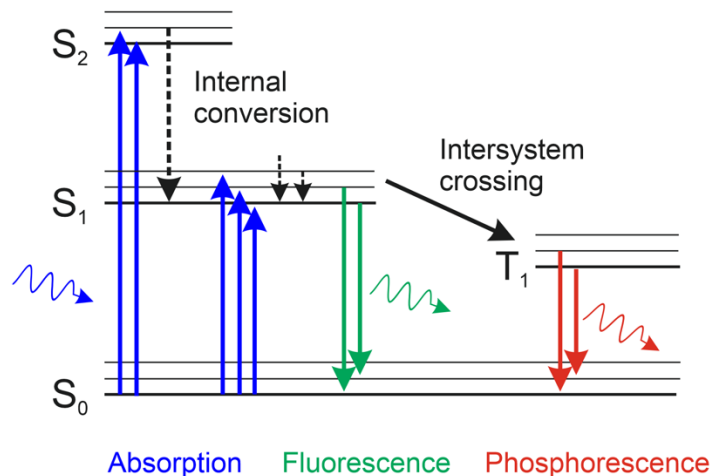
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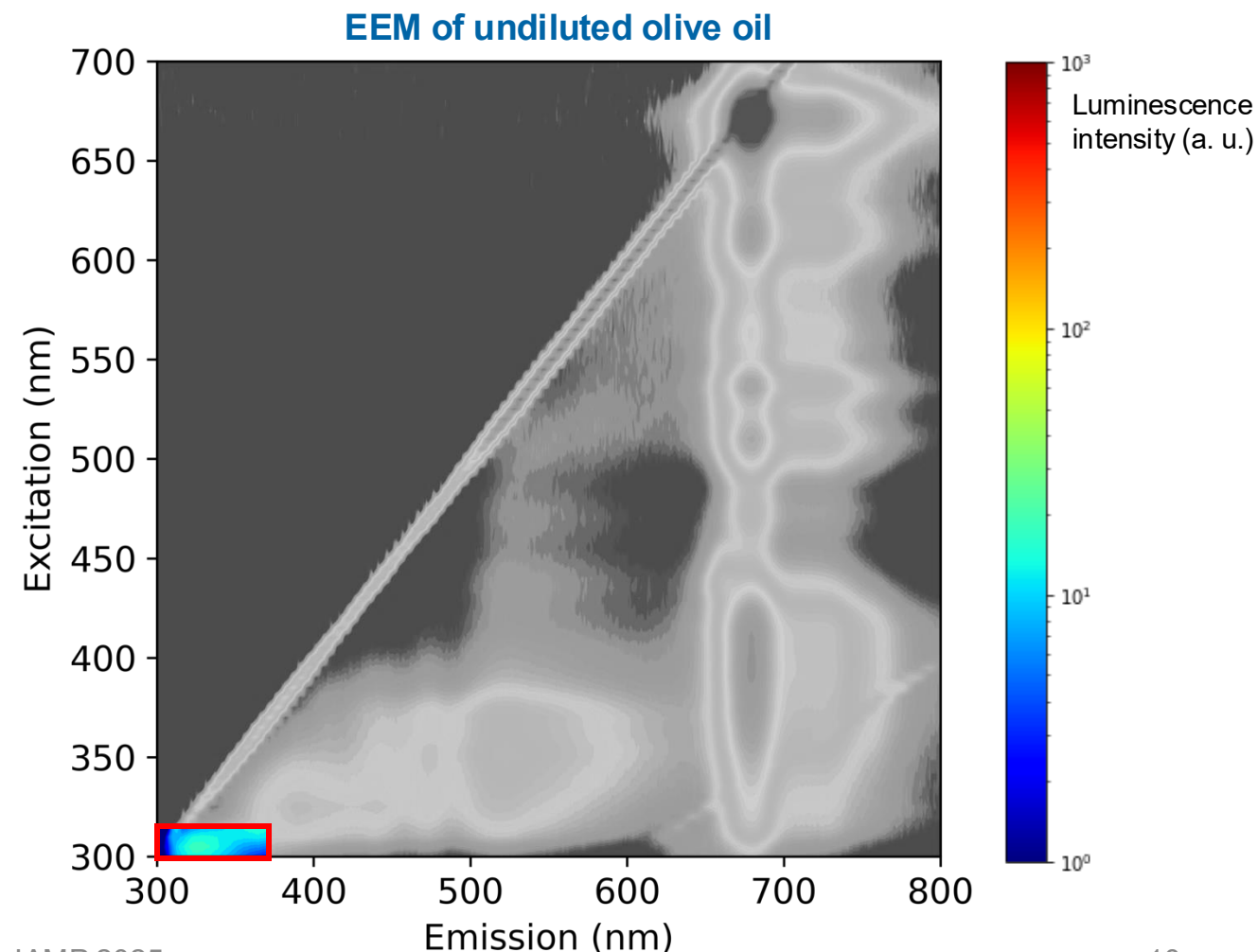
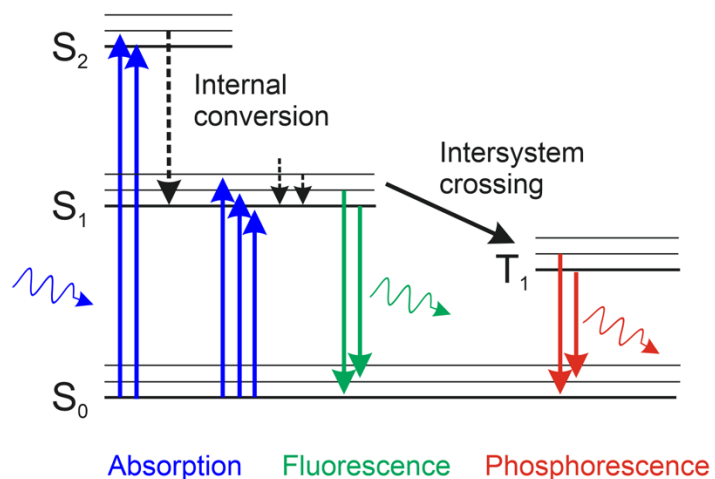
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Method: fluorescence

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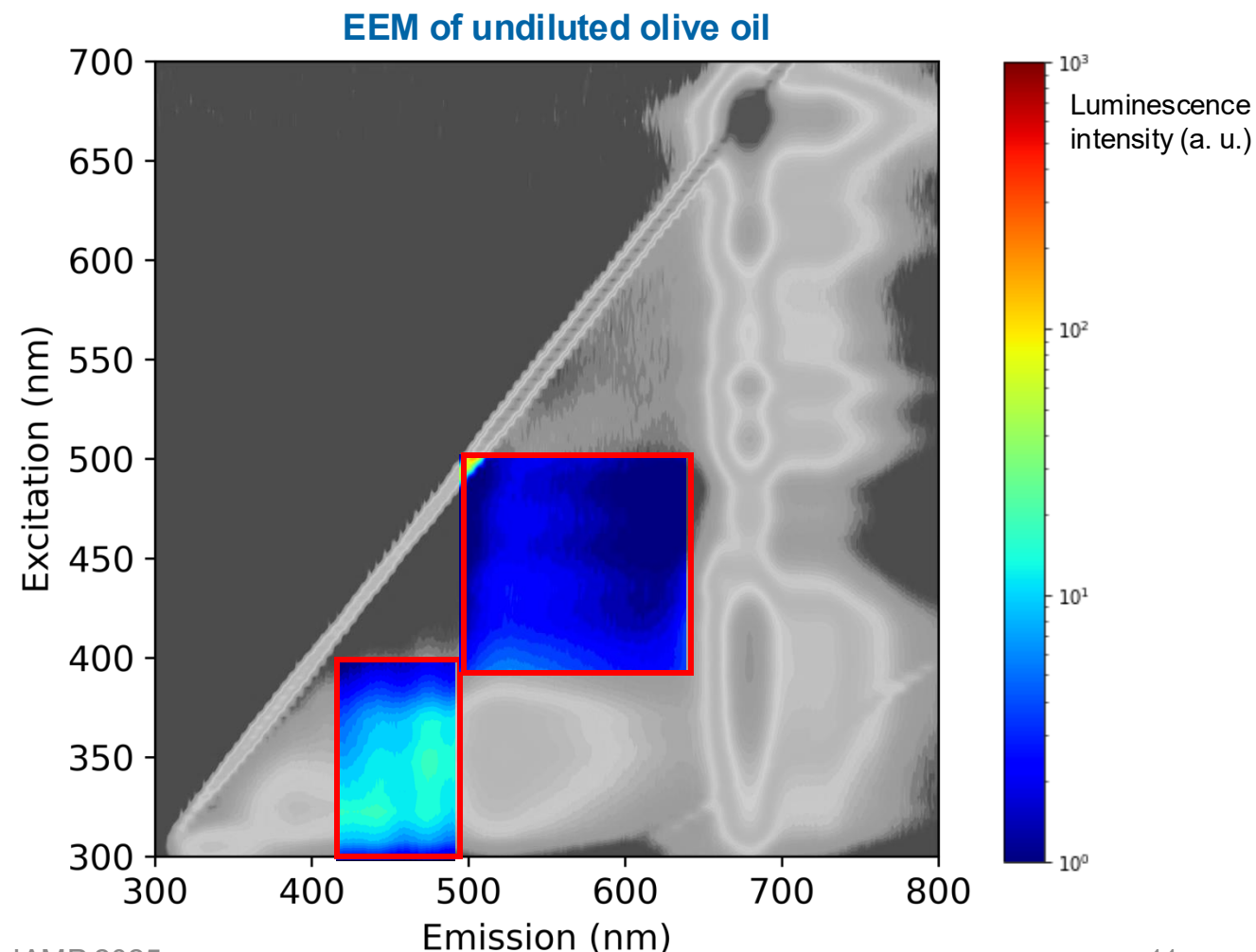


Method: fluorescence

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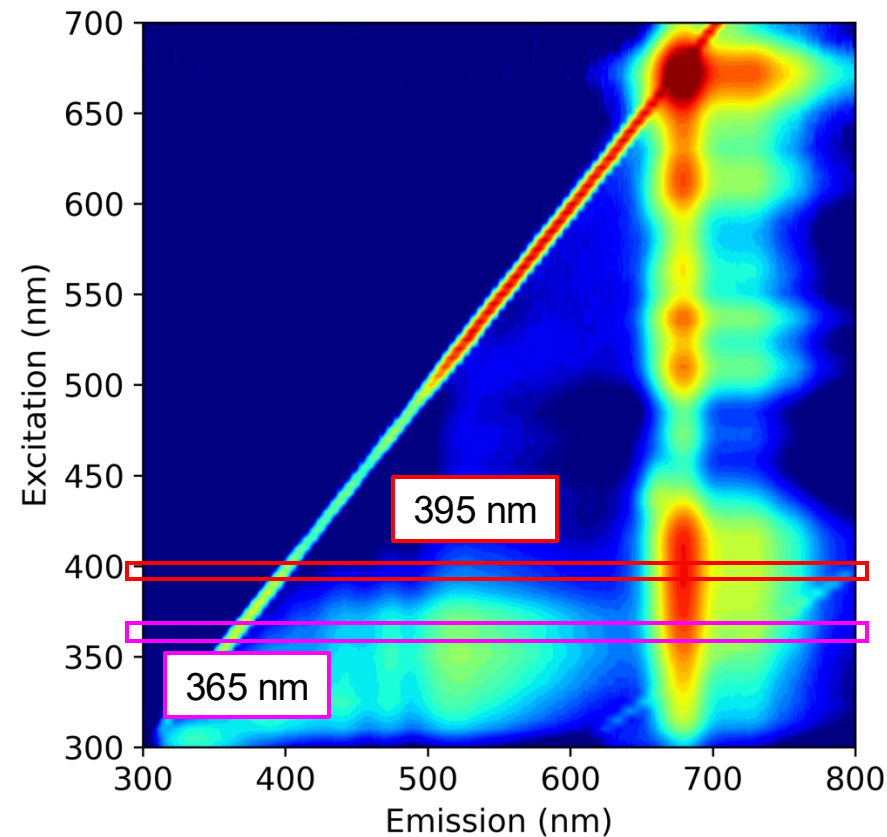
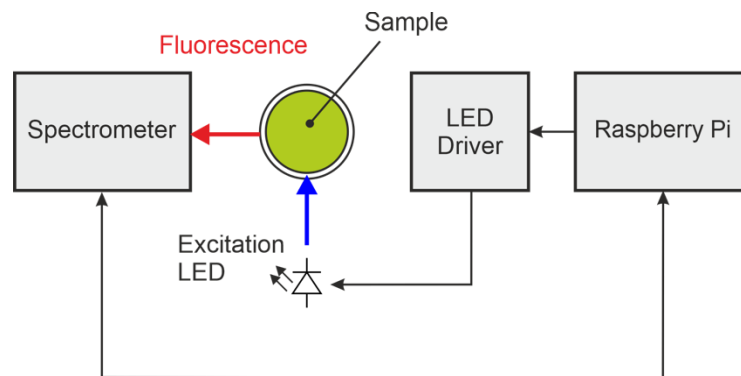
- Chlorophylls *a*, *b*, *d*
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➤ Fluorescence is very specific of the molecules contained in the sample
→ Fluorescence can be used for quantitative assessment



Extraction of quality parameters with fluorescence: instrument

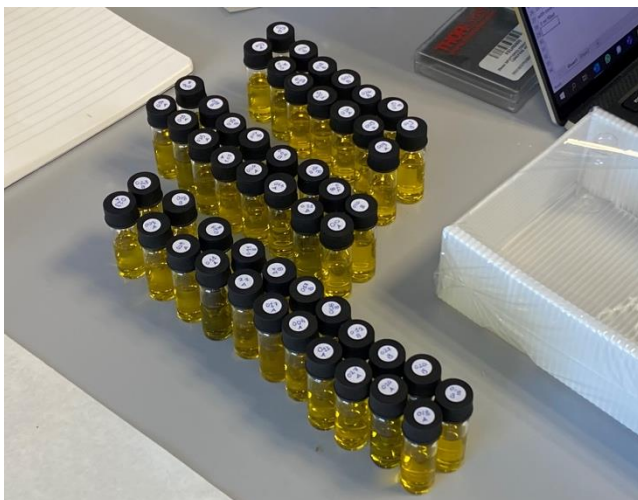
- Single spectra
- Compact and cost-effective design
- No optical filters, lenses
- Fast acquisition (<1s)
- Easy to use



Venturini, F. et al., *Foods* 2021, 10, 1010

Dataset of olive oils

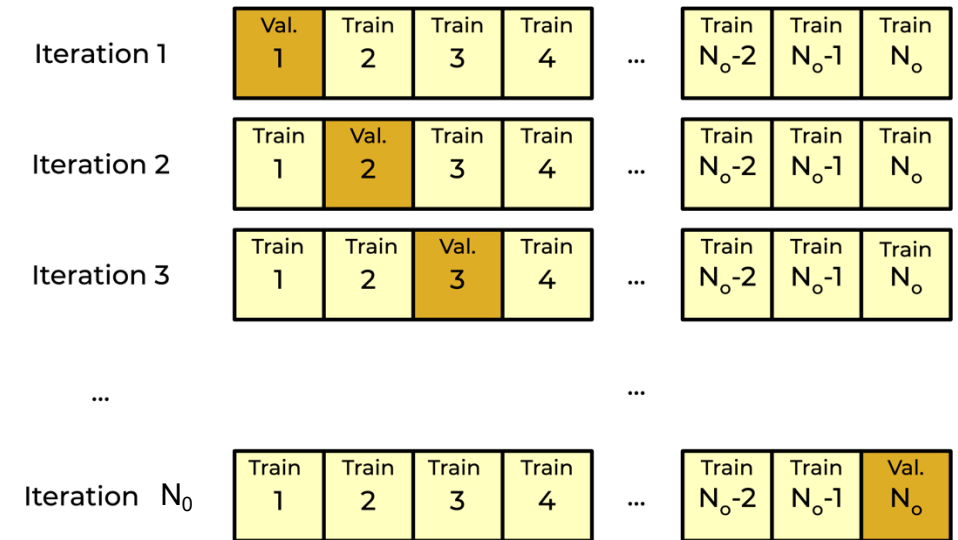
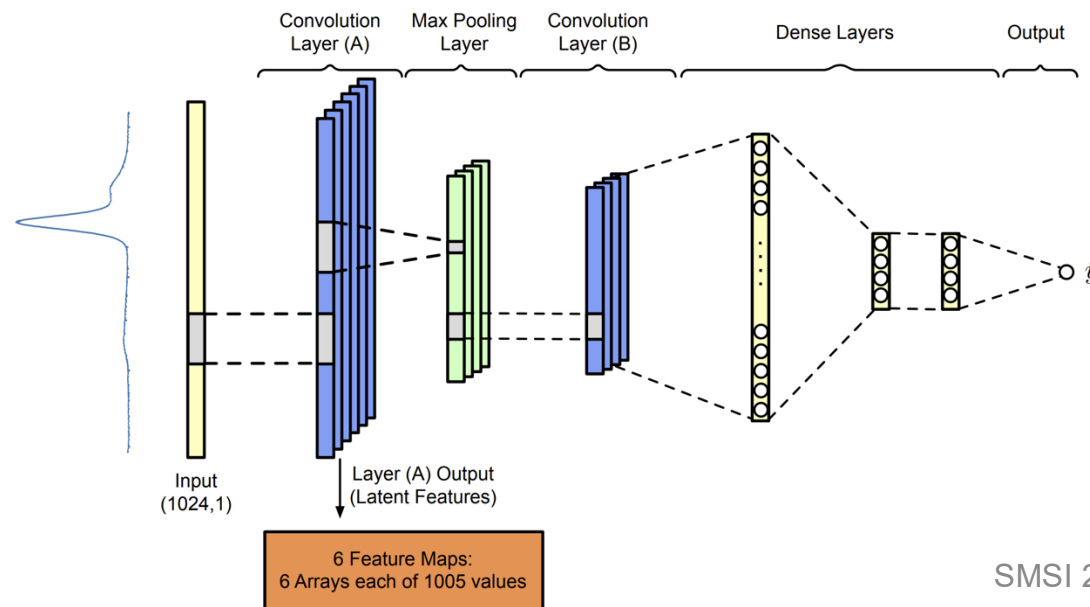
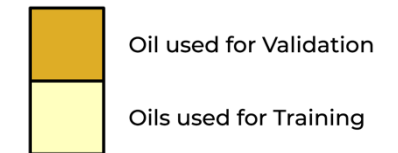
- Oils from single producer, 3 different quality levels: EVOO, VOO, LOO
- Complete chemical analysis for all oils



Label	Acidity (%)	Peroxide value (mEq O ₂ /kg)	K ₂₇₀	K ₂₃₂	Ethyl esters (mg/Kg)	Quality
D03	0.35	8.4	0.123	1.435	26	VOO
D04	0.34	8.6	0.108	1.403	40	VOO
D05	0.36	10.3	0.112	1.44	18	VOO
D06	0.31	9.2	0.151	1.484	18	VOO
D07	0.50	8.9	0.150	1.537	47	VOO
D08	0.40	8.5	0.158	1.546	25	VOO
D19	0.25	4.9	0.13	1.540	10	EVOO
D20	0.26	4.6	0.14	1.540	10	EVOO
D35	0.17	6.4	0.12	1.63	8	EVOO
D38	0.16	6.4	0.12	1.63	9	EVOO
D45	0.17	4.9	0.12	1.63	7	EVOO
D46	0.18	5.0	0.13	1.63	8	EVOO
D47	0.18	5.2	0.13	1.64	16	EVOO
D49	0.9	9.9	-	-	-	LOO
D51	2.16	-	-	-	-	LOO
D52	1.78	22	-	-	-	LOO
D53	0.7	8.7	-	-	-	LOO
D64	0.2	7.1	0.13	1.63	29	VOO
D73	0.2	8.9	0.14	1.66	15	EVOO
D77	0.24	10.4	0.13	1.74	26	VOO
D81	0.16	4.9	0.12	1.63	9	EVOO
D92	0.18	5	0.17	1.91	15	EVOO

Design of the 1D-CNN

- Design using **physical** information:
 - Number of filters $\rightarrow$ number of the expected features
 - Size of filters $\rightarrow$ resolution of the spectrometer
- **Supervised** learning
- Cross validation $\rightarrow$ **Leave-one-out**



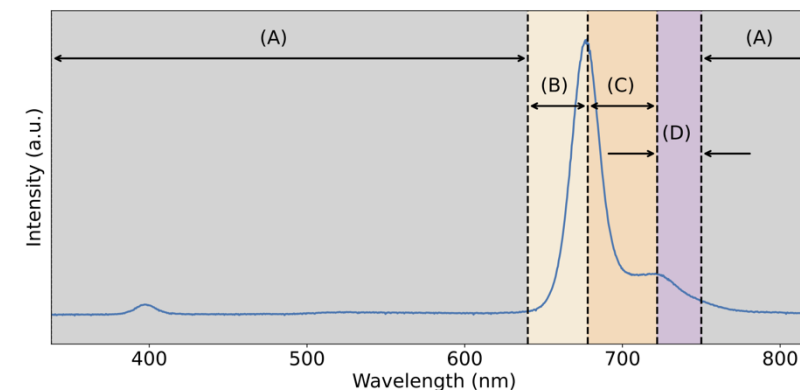
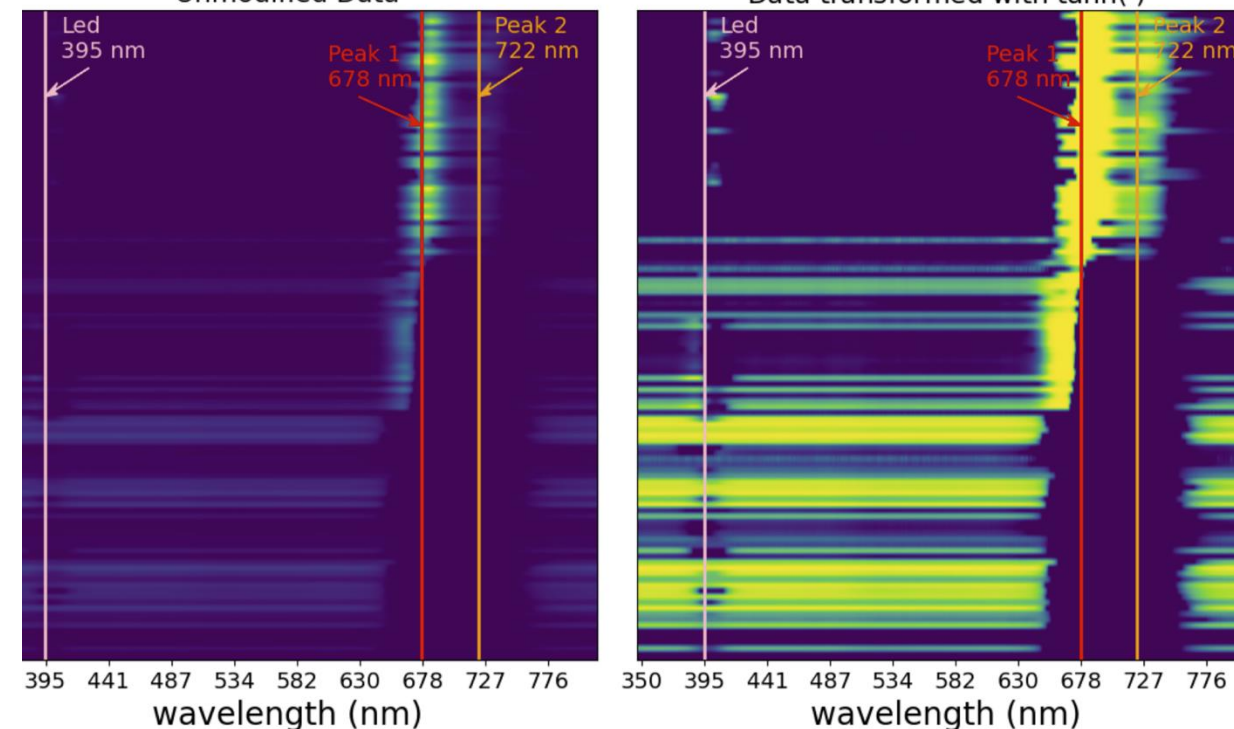
Explainability results

Feature maps for all the oils

K_{232}

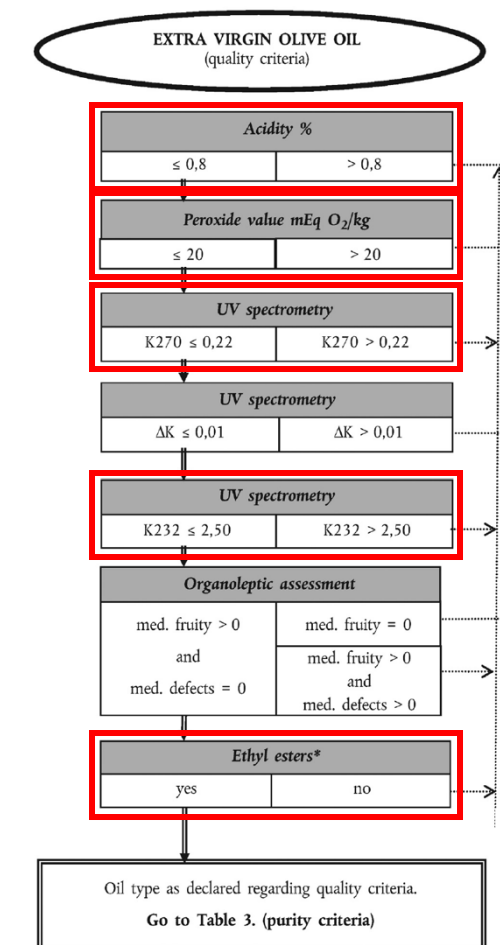
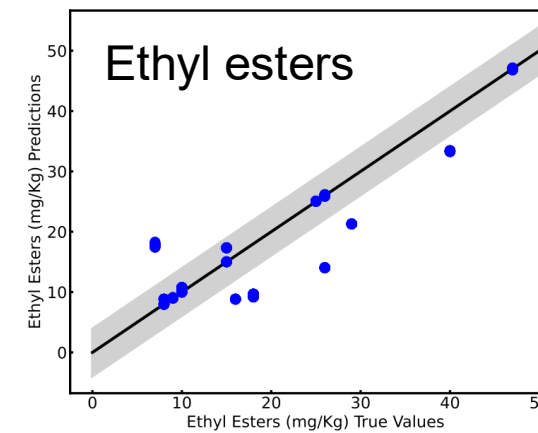
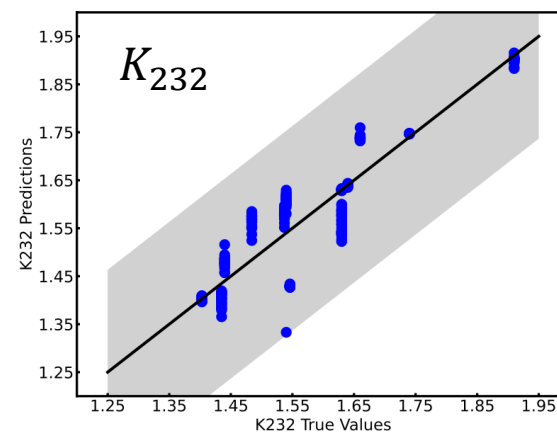
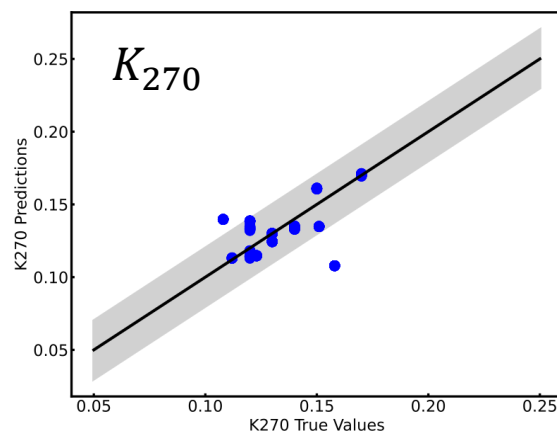
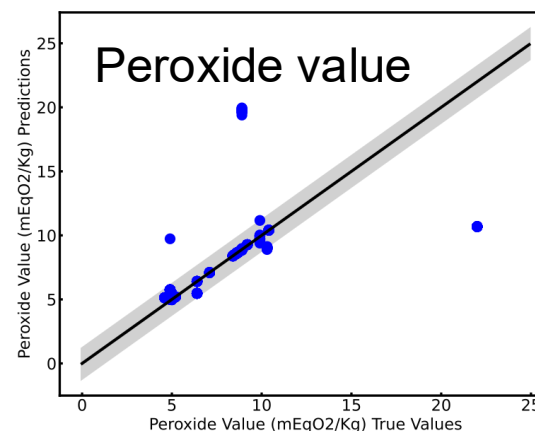
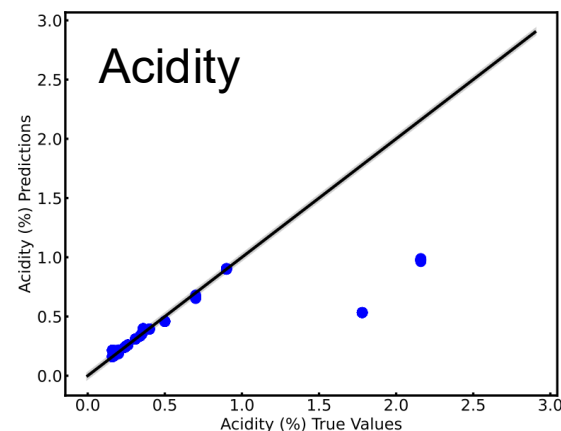
Unmodified Data

Data transformed with $\tanh(\cdot)$



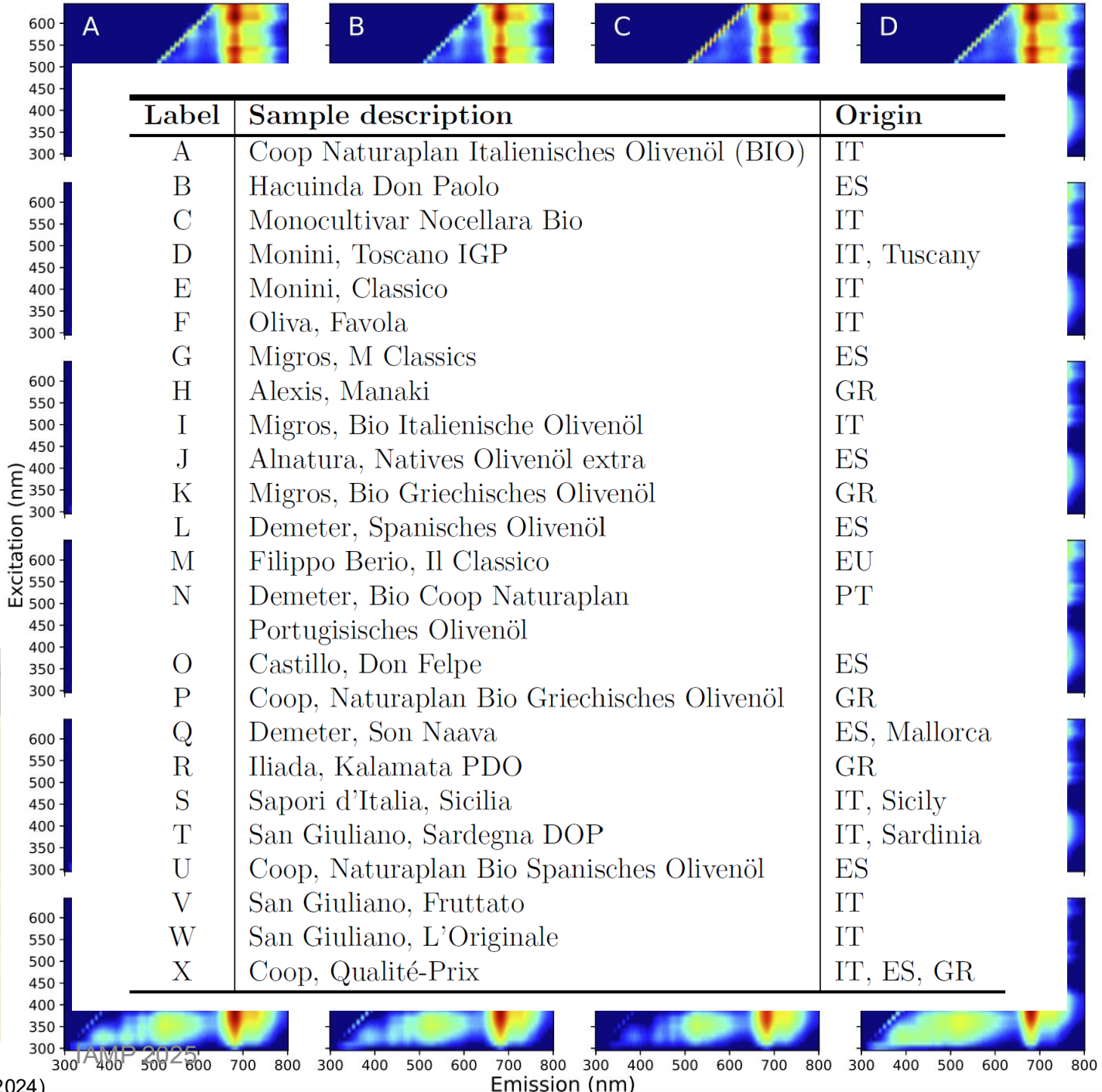
- There 1D-CNN identifies the **spectral regions** relevant for the prediction
- The 1D-CNN learns from the **same** regions of the spectra for **all oils** → results are **robust**
- The 1D-CNN identifies the spectral features which are physically meaningful → **trust** the results

Results: extraction of quality parameters



Ageing of EVOO

- 24 commercial EVOO
- Heterogenous geographical origin (Italy, Spain, Portugal, Greece, and unspecified Europe) and price
- 3 samples per oil and ageing step
- 10 ageing steps at 60°C, dark, sealed vials
- 720 EEMs



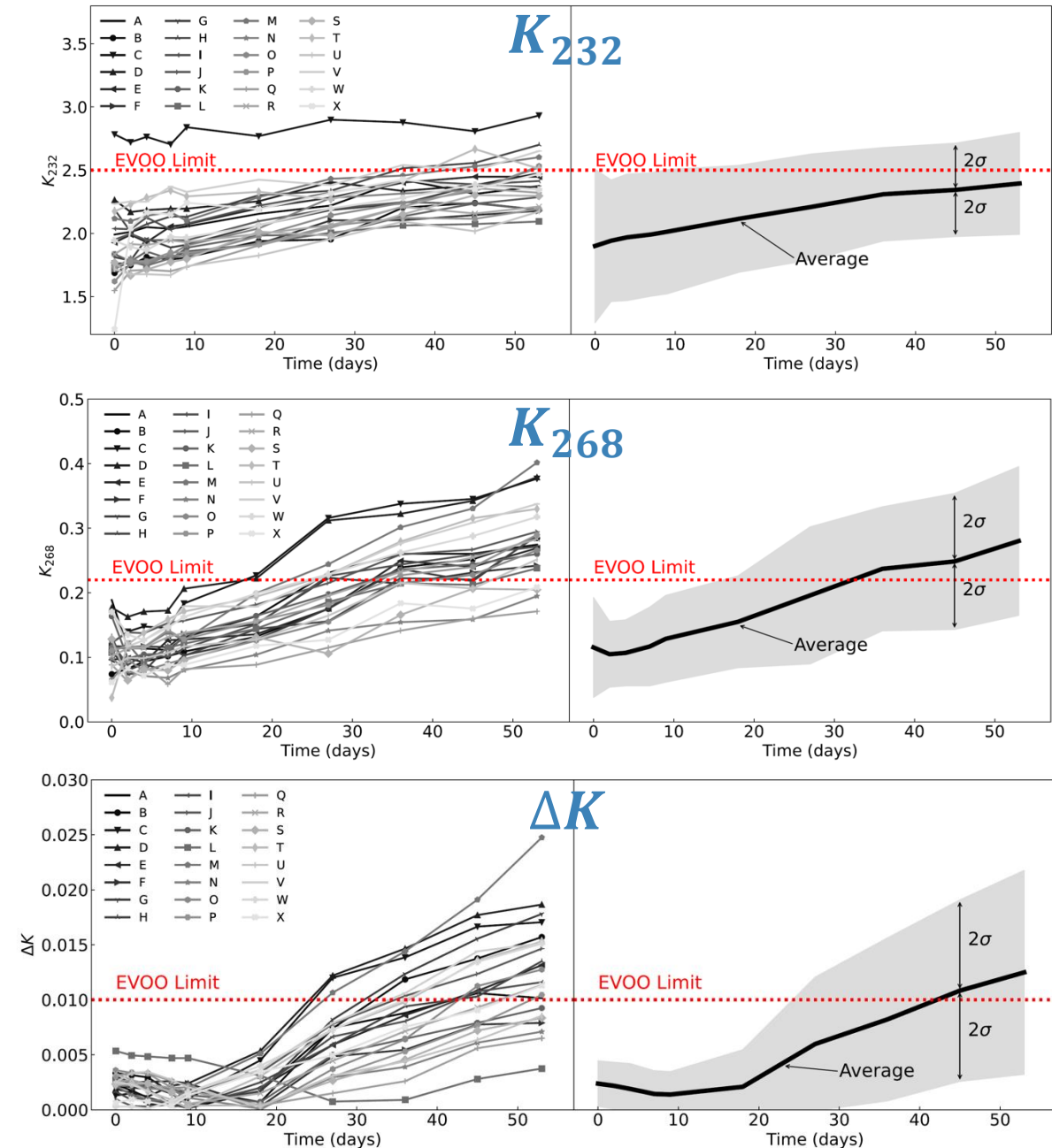
Quantitative assessment of ageing: UV spectroscopy

- Ageing assessed through quality parameters: K_{232} , K_{268} , ΔK

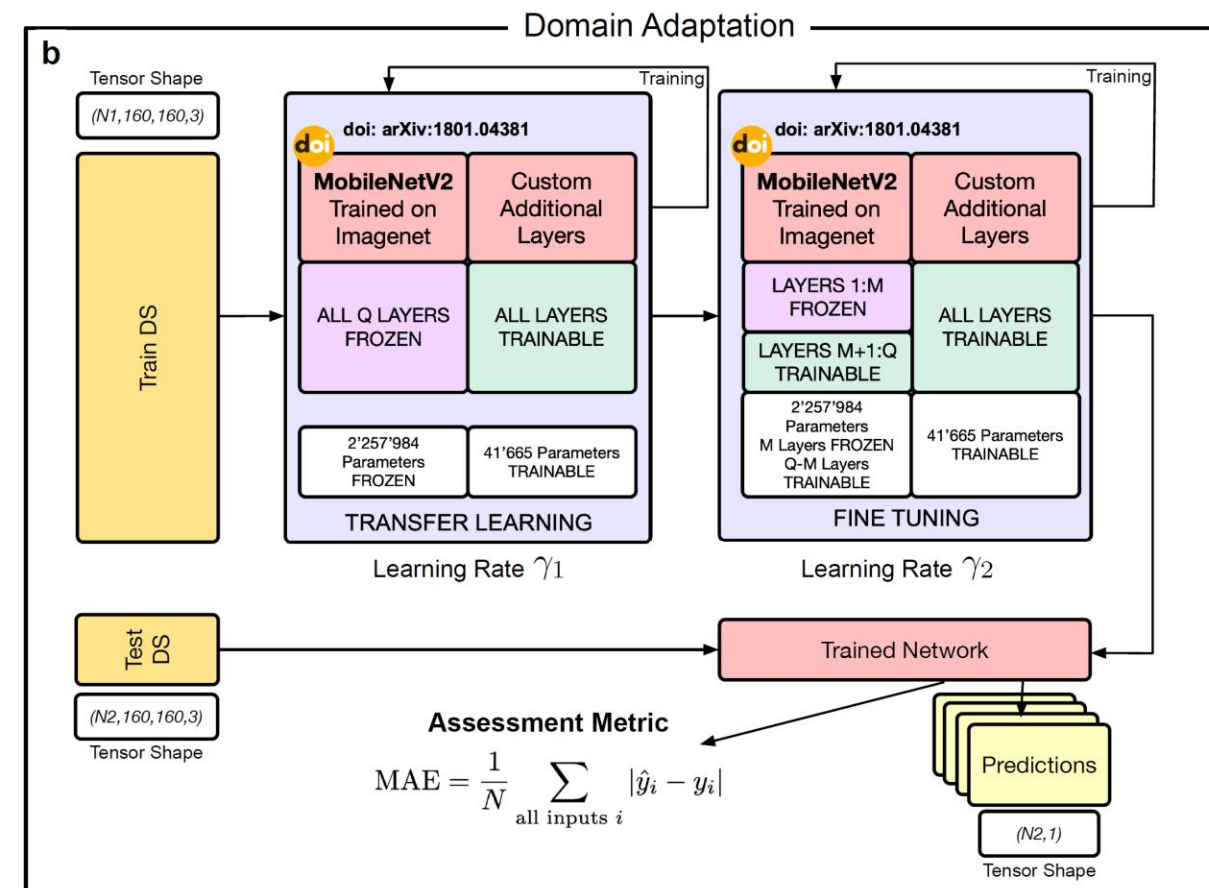
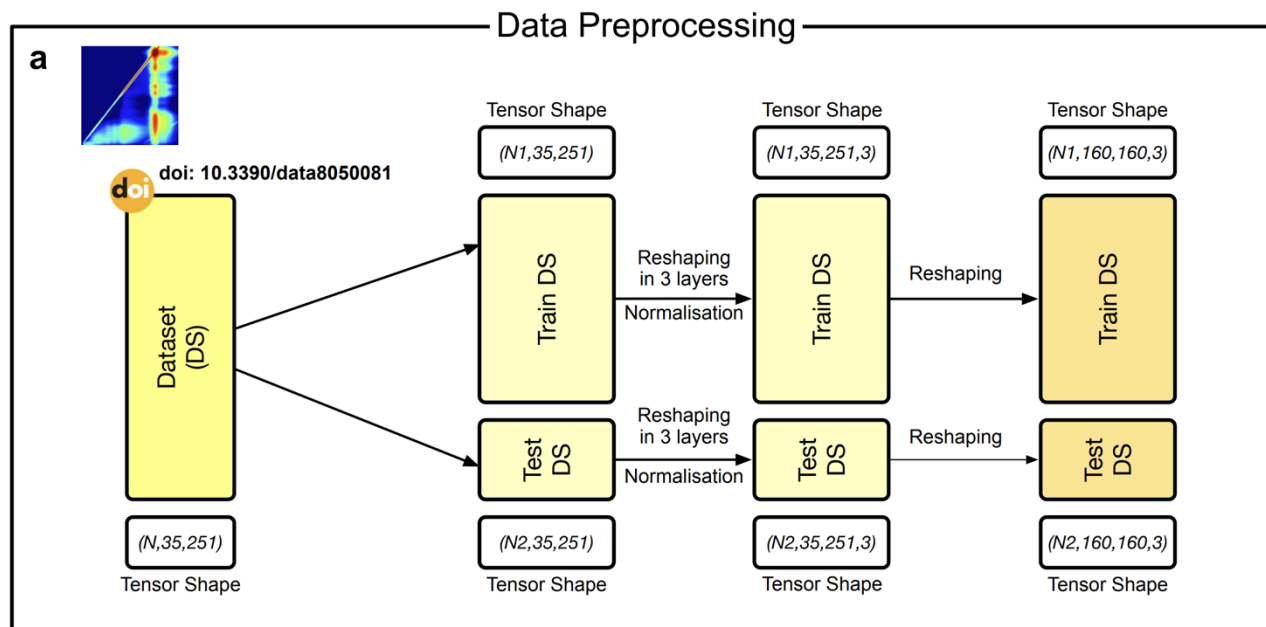
Commission Regulation EEC 2568/91,
Commission Implementing Regulation (EU) No 1348/2013

Observations:

- Similar trends for all oils despite heterogeneity
- K_{232} : total change 26% → weak
- K_{268} : total change 143% → strong
- ΔK : Behaviour similar to K_{268}

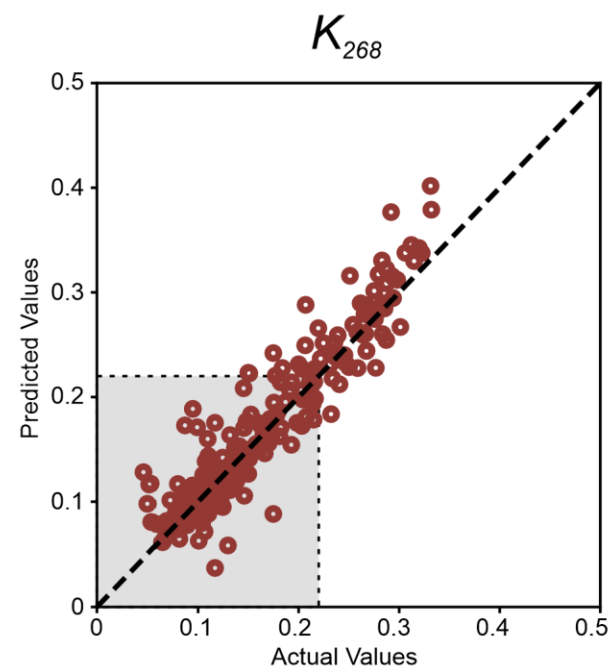
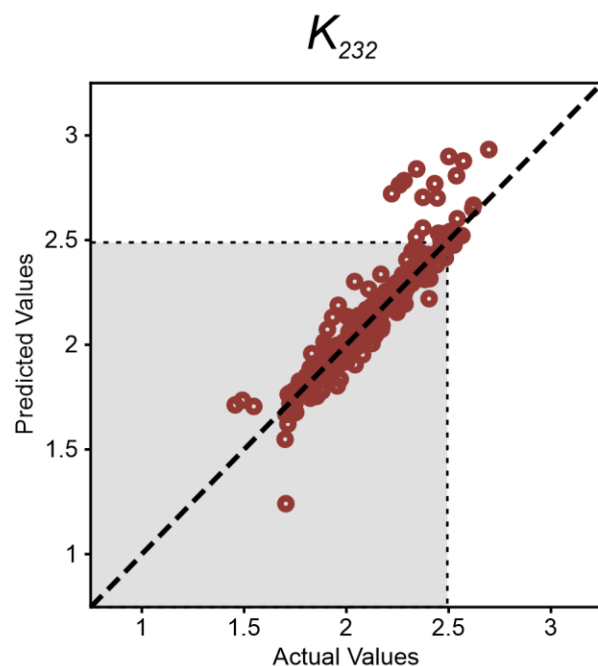


Deep learning approach



Michelucci, U., Venturini, F. *Sci Rep* **14**, 22291 (2024)

Results: extraction of quality parameters

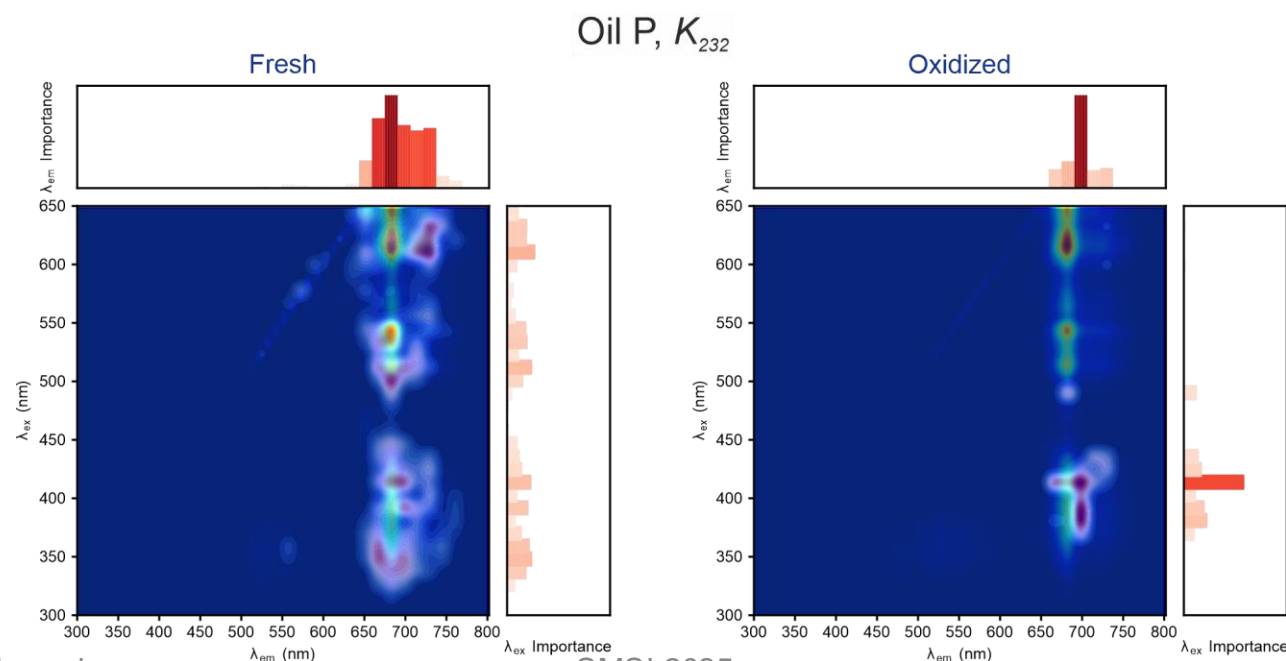
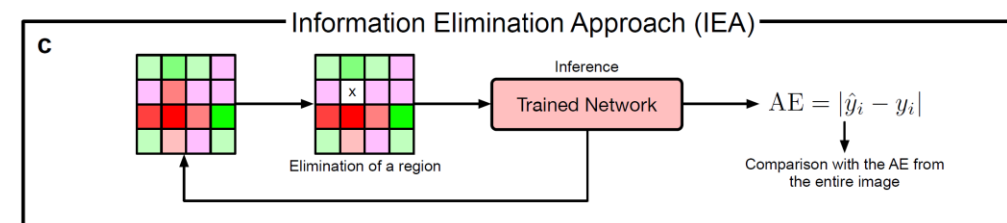


Quality parameter	MAE	Label error	EU limits
K232	0.066	0.06	2.5
K268	0.010	0.02	0.22

➤ Fluorescence intensity can be used to predict UV quality parameters

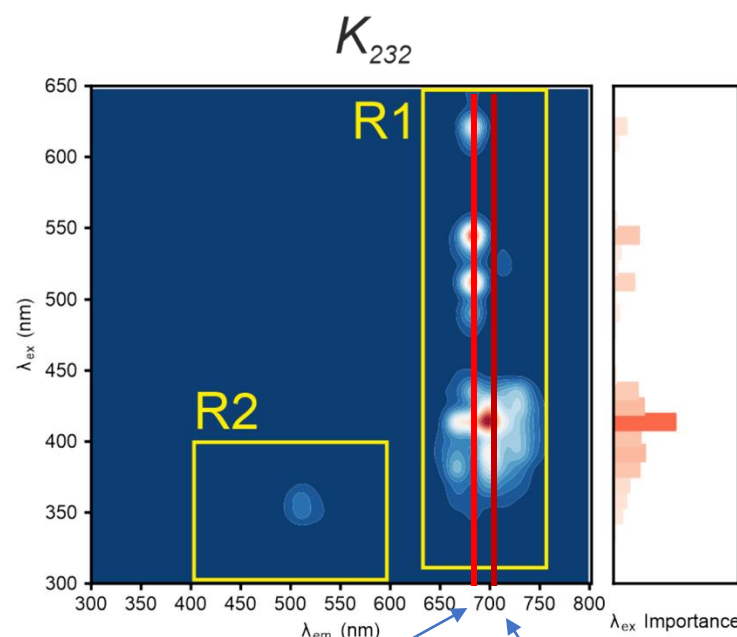
Explainability approach

- Method: Information elimination approach
- Heatmap of the AE (absolute error) when eliminating a spectral region
→ identification of spectral regions contributing to the prediction

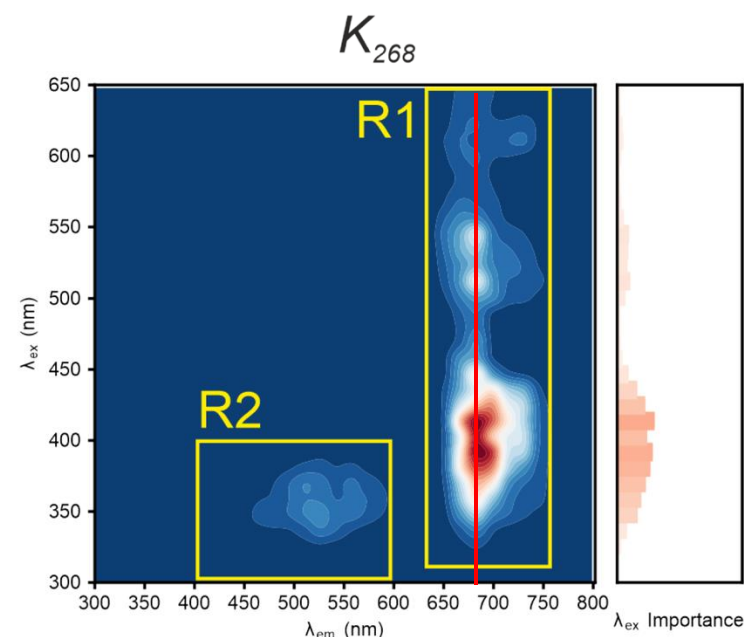


Results: heatmap of the spectral regions relevant for prediction

- Identification of spectral regions affected during the ageing (the oxidation process)



R1: chlorophylls → pheophytins



R1: reduction conc chlorophylls
R2: reduction conc carotenoids

No free lunch: challenges in application of ML

