

Data augmentation for pathogen segmentation in vinewood fluorescence microscopy images

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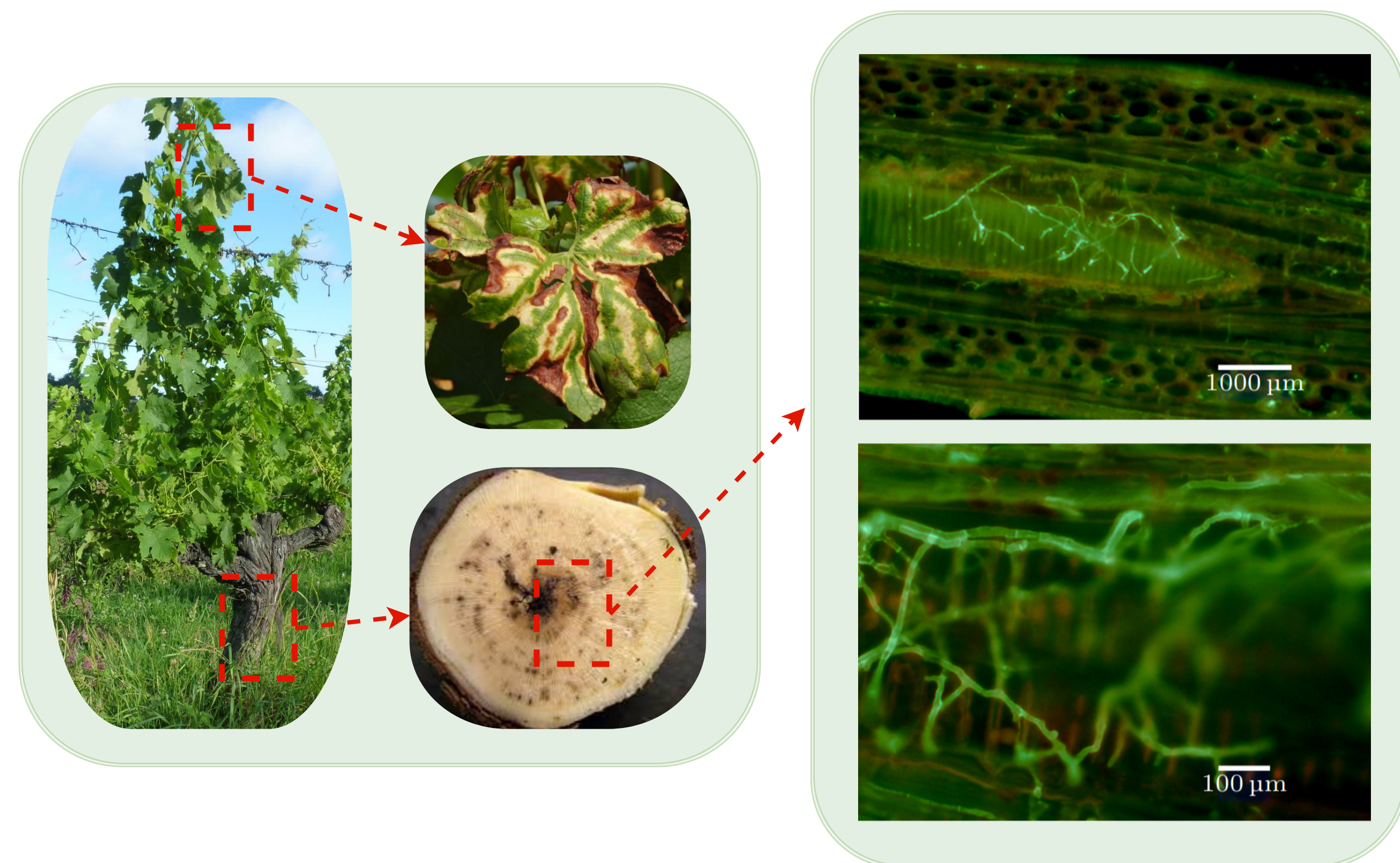
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MOTIVATION-PROBLEMS

Grapevine trunk diseases pose a major problem for vinegrowers worldwide.

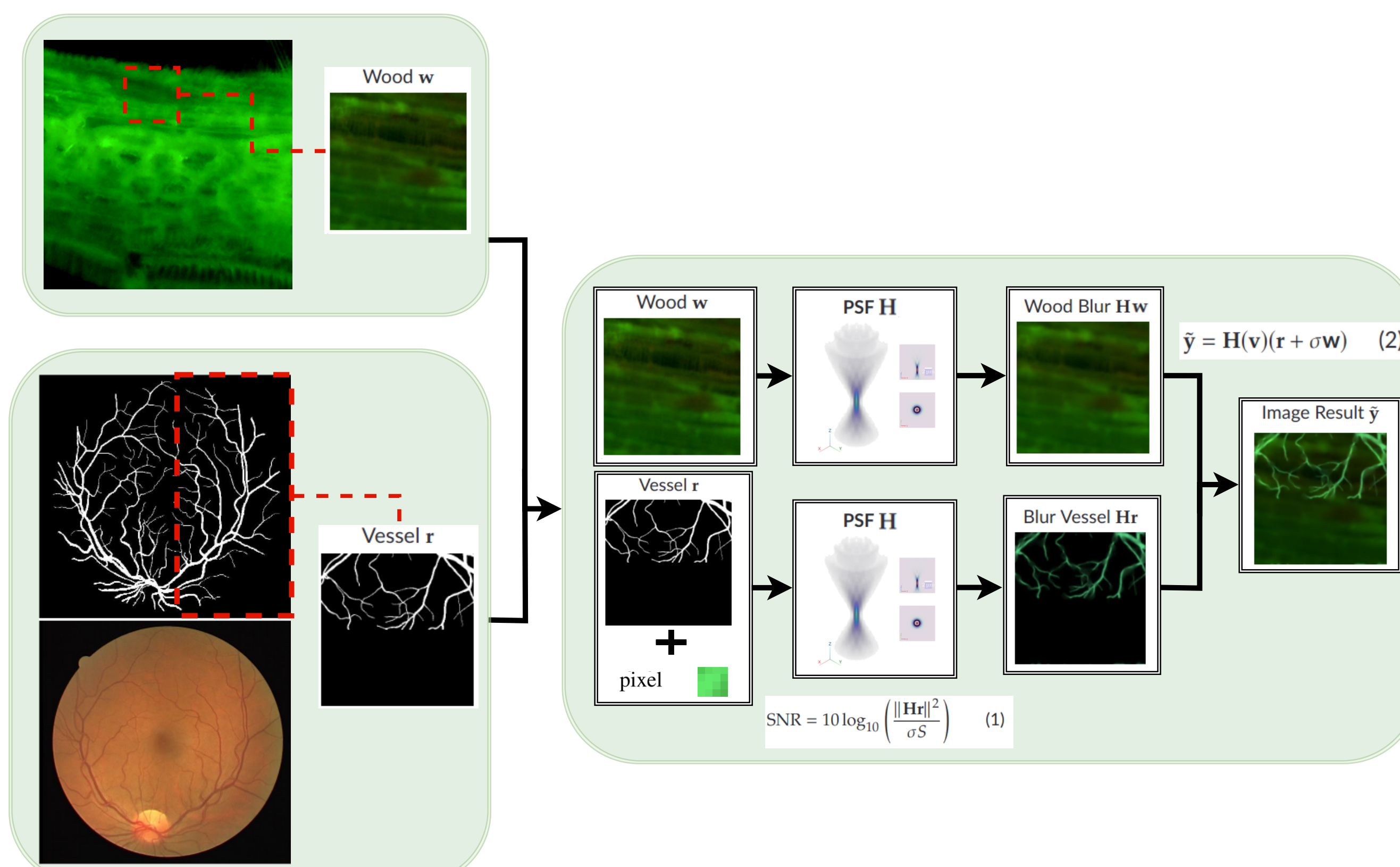
- 13% of vineyards are unproductive each year in France, resulting in important commercial losses [1].
- We are studying one of the oldest diseases: **Esca**.
- **No treatment exists** once the wood is infected.
- To study the behavior of pathogens, which is still **poorly understood**, we make inoculation experiments and use **fluorescence microscopy**.



Getting this type of image is challenging, time-consuming, and our collection is limited.

SYNTHETIC IMAGE GENERATION

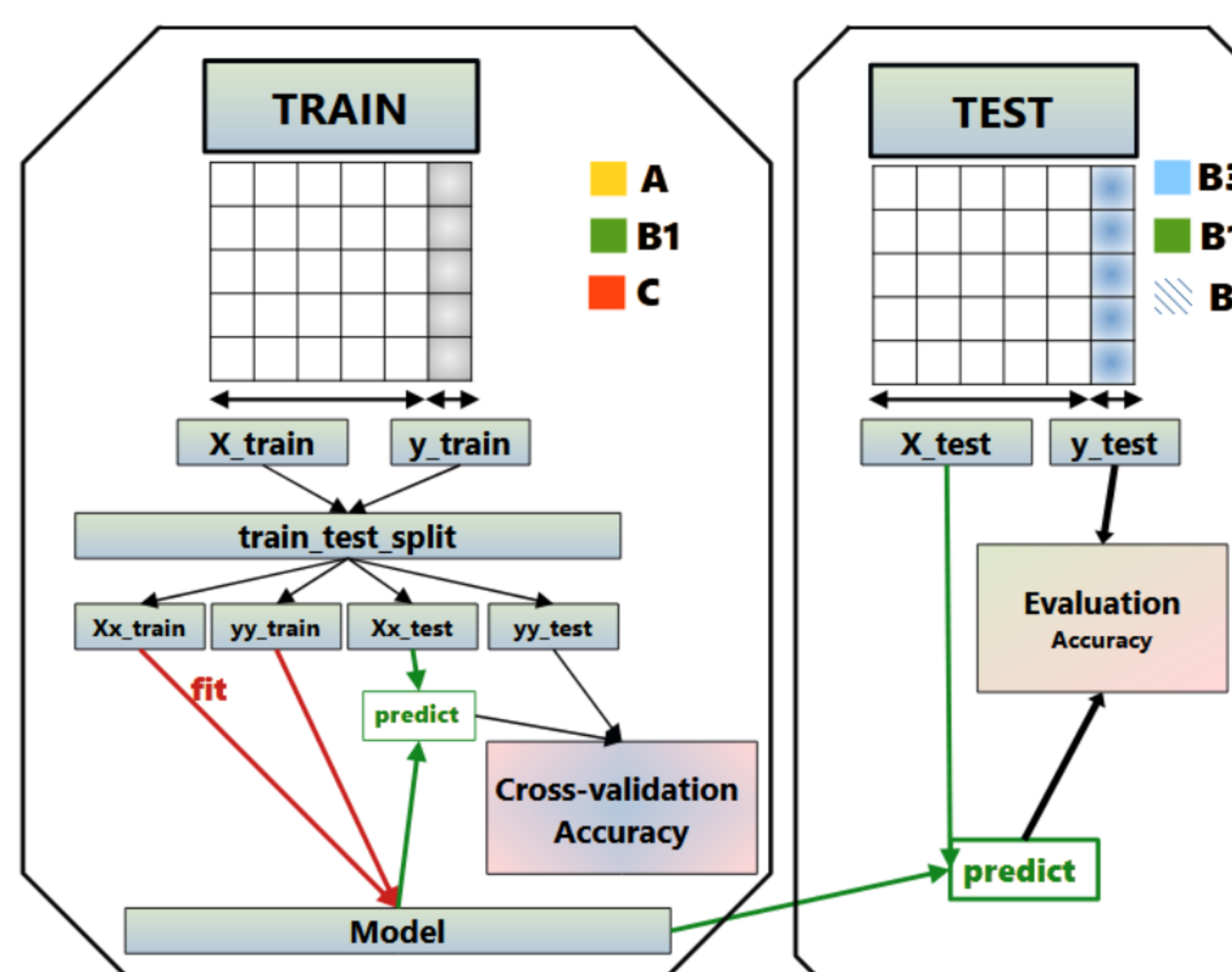
Our objective: offset the deficiency of images by creating realistic images containing the desired filamentary pattern and variable blur effect and employing a data augmentation approach.



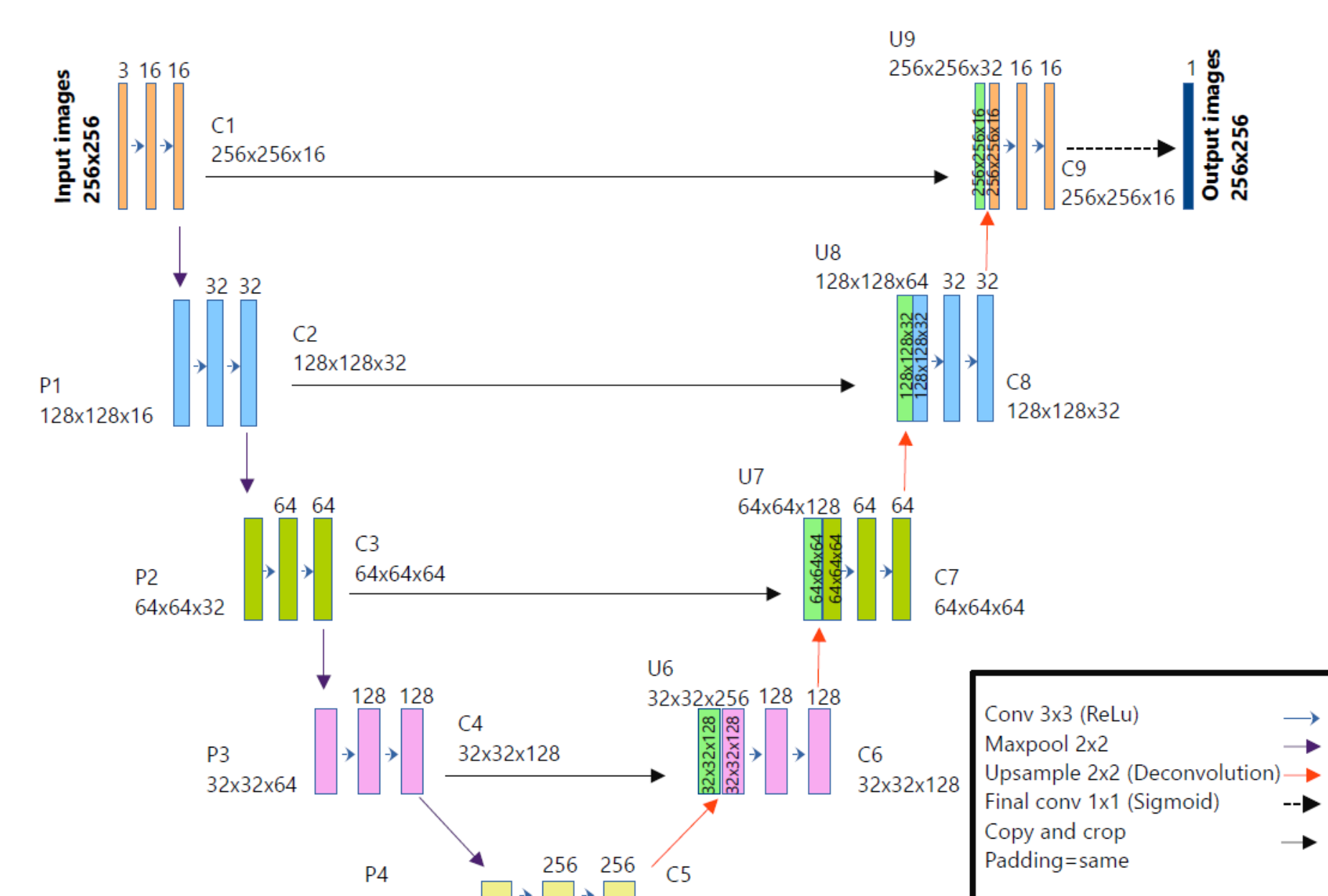
AVAILABLE DATASETS

dataset name	B: real images				
	A	B1	B2	B3	C
description	synthetic images	with fungi in lower quality image	without fungi	with fungi	mix A, B1 and B3
number of images	427	247	312	569	128
images					
segmentation					

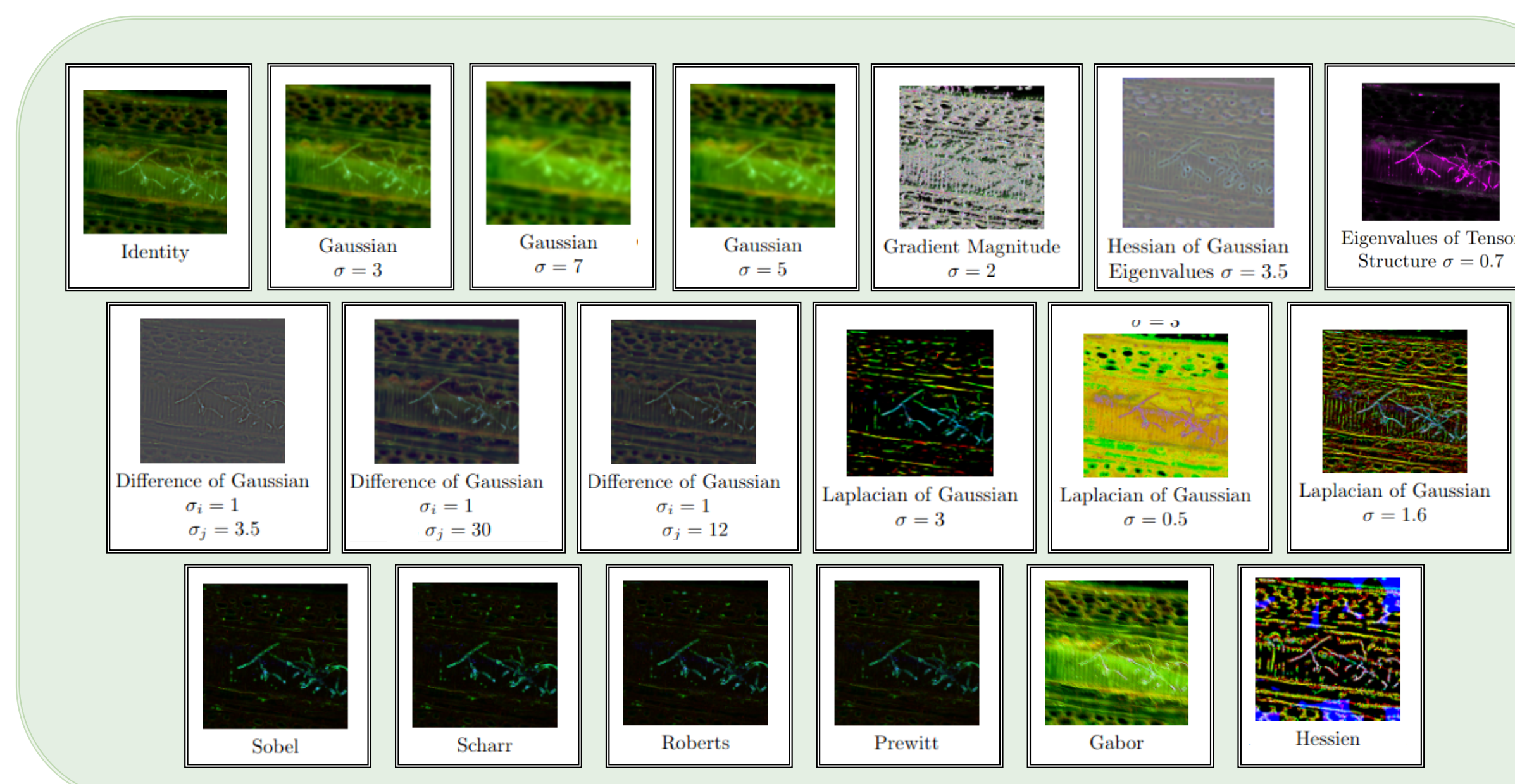
TRAINING & TEST



LEARNING WITH U-NET [2]



LEARNING WITH RANDOM FOREST [3]

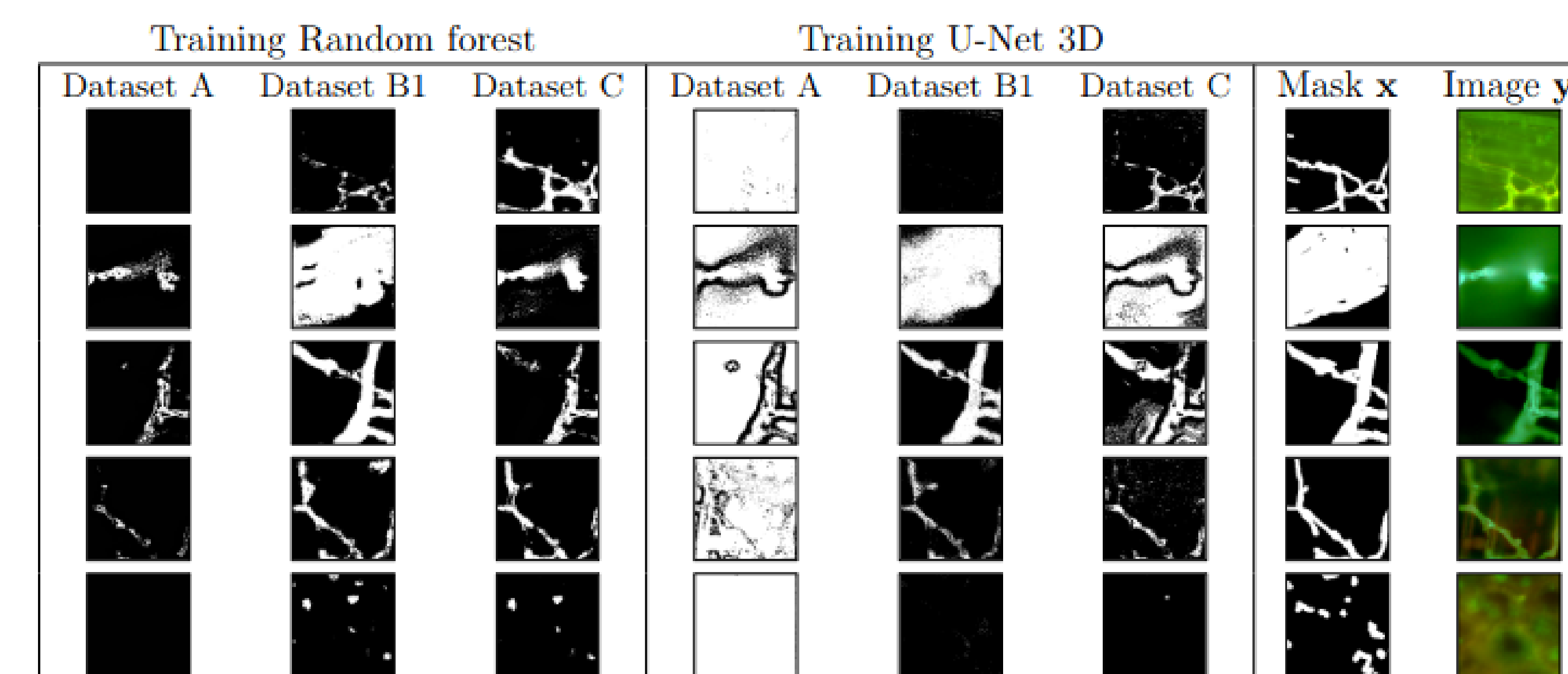


RESULTS

Segmentation accuracies, as a function of the segmentation method, the training database, and the testing database

Training database:	Random forest			U-Net 3D		
	Dataset A	Dataset B1	Dataset C	Dataset A	Dataset B1	Dataset C
CV accuracy	92.96%	96.16%	97.06%	89.68%	93.84%	90.96%
Dataset B1	2.69%	94.13%	95.70%	18.98%	90.10%	83.25%
Dataset B	41.00%	95.70%	92.45%	10.47%	94.17%	85.73%
Dataset B3	6.58%	92.14%	88.08%	10.83%	93.79%	89.00%

Example of results on real images



- This data augmentation based on image formation helps segmenting lower-quality images.
- This method could be generalized to other microscopy imaging techniques.

REFERENCES

- [1] Lecomte, P. et al. (2018). Esca of grapevine and training practices in France. *Phytopathologia mediterranea*, 57(3), 472-487.
- [2] Ronneberger, O. et al. (2015). U-net: Convolutional networks for biomedical image segmentation. In *MICCAI 2015*, pp. 234-241.
- [3] Breiman, L. (2001). Random forests. *Machine learning*, 45, 5-32