



Pole for Doctoral Studies
Center for Doctoral Studies Sciences and Techniques and Medical Sciences

ANNOUNCEMENT OF DOCTORAL THESIS DEFENSE



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**Will present their research work with the aim of earning a
Doctorate**

**Doctoral program: Engineering Sciences and Techniques
Discipline: Engineering Sciences and Techniques
Specialty: Embedded Systems and Artificial Intelligence**

**On 11/07/2026 at 11H00 at the Conference Hall, Polydisciplinary
Faculty of Larache, UAE
Under the Theme**

**Intelligent Embedded Systems for Precision Floriculture:
Knowledge Distillation and Architecture Optimization for Real-
Time Edge Deployment in Moroccan Rosa Damascena
Monitoring**

Front of the jury composed of :

First Name & Last Name	Establishment	Designation
Pr. SETTI Larbi	FP of Larache, UAE	President
Pr. ACHKARI BEGDOURI Mohammed	FP of Larache, UAE	Reviewer
Pr. LAGMICH Youssef	FP of Larache, UAE	Reviewer
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Pr. EL ABDERRAHMANI Abdellatif	FS of Fez, USMBA	Examiner
Pr. OULHAJ Otman	ENS of Tetouan, UAE	Examiner
Pr. LASRI Rafik	FP of Larache, UAE	Supervisor

Host Research Structure: Technologies Émergentes Durables : Agriculture, Environnement et Énergies Propres

Abstract



Rosa damascena, the Damask rose, is a central crop of the Dades Oasis in southeastern Morocco, where farmers around Kelaat M’Gouna cultivate thousands of kilometres of hedges for essential oil and rose-water production. Maturity assessment and yield estimation are still carried out by eye, and no automated tool supports them. State-of-the-art deep learning methods exist, but their hardware requirements—GPU workstations or cloud services—are out of reach in this setting, where the realistic deployment platform is a Raspberry Pi 4B. This thesis develops and deploys four lightweight deep-learning methods that bring accurate real-time monitoring onto that hardware.

The work follows a thesis-by-publication format and is built around four peer-reviewed publications, evaluated on a single field collected dataset from the Dades Oasis (3,114 maturity-labelled images; 2,807 annotated detection images) and benchmarked on the same Raspberry Pi 4B. ATMS-KD is an adaptive knowledge-distillation framework whose compact student reaches 97.11% accuracy at 72.19 ms per image, exceeding eleven distillation baselines by 1.60 pp. YOLOv10-N, fine-tuned on rose imagery, achieves 92.12% precision and 90.65% mAP@0.5, and is coupled with a statistical sampling procedure that predicts field-level yield within 3.2% of actual harvest.

GWOEdgeNet uses the Grey Wolf Optimizer to design an ultra-lightweight CNN (10,946 parameters); trained with decoupled distillation, it reaches $95.15\% \pm 0.96$ accuracy at 38.1 ms (26.2 FPS)—the only model in an eight baseline comparison that crosses the 24 FPS real-time threshold on the target hardware. Wavelet Scattering replaces the learned front-end with a frozen Morlet transform and trains only a 61,410-parameter CNN head, reaching 96.0% accuracy and $F1 = 0.96$ with 5.6 to 383 times fewer trainable parameters than six ImageNet-pretrained baselines. The thesis shows that accurate deep-learning monitoring can be deployed on affordable hardware under the specific conditions of the Dades Oasis.

Keywords : Knowledge distillation; decoupled knowledge distillation; neural architecture search; Grey Wolf Optimizer; YOLOv10; Rosa damascena; edge AI; embedded systems; Raspberry Pi 4B; TinyML; Dades Oasis