

IntelliSys²⁰²²

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How to Improve the Teaching of Computational Machine Learning Applied to Large-Scale Data Science

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What is the Problem?



In public universities in Mexico, specifically in ENES Morelia UNAM, Machine Learning (ML), Deep Learning (DL), and Big Data subjects are taught in the curriculum plan.

These subjects are taken from the 4th semester up to the 8th semester. Nevertheless, there are some cases where the students cannot identify and solve AI, ML, DL, and Big Data problems

Solution Proposal

Modification and improvement to the courses that adhere to the preferred requirements of the students. The solution to these problems is to give higher priority to practice with real data and use cases than to abstract theory

How to Improve Teaching

We implemented a previous knowledge survey, where the students and teachers select the topics of interest.

To solve the learning deficit, we propose a ML course based on the more unfamiliar and requested topics.



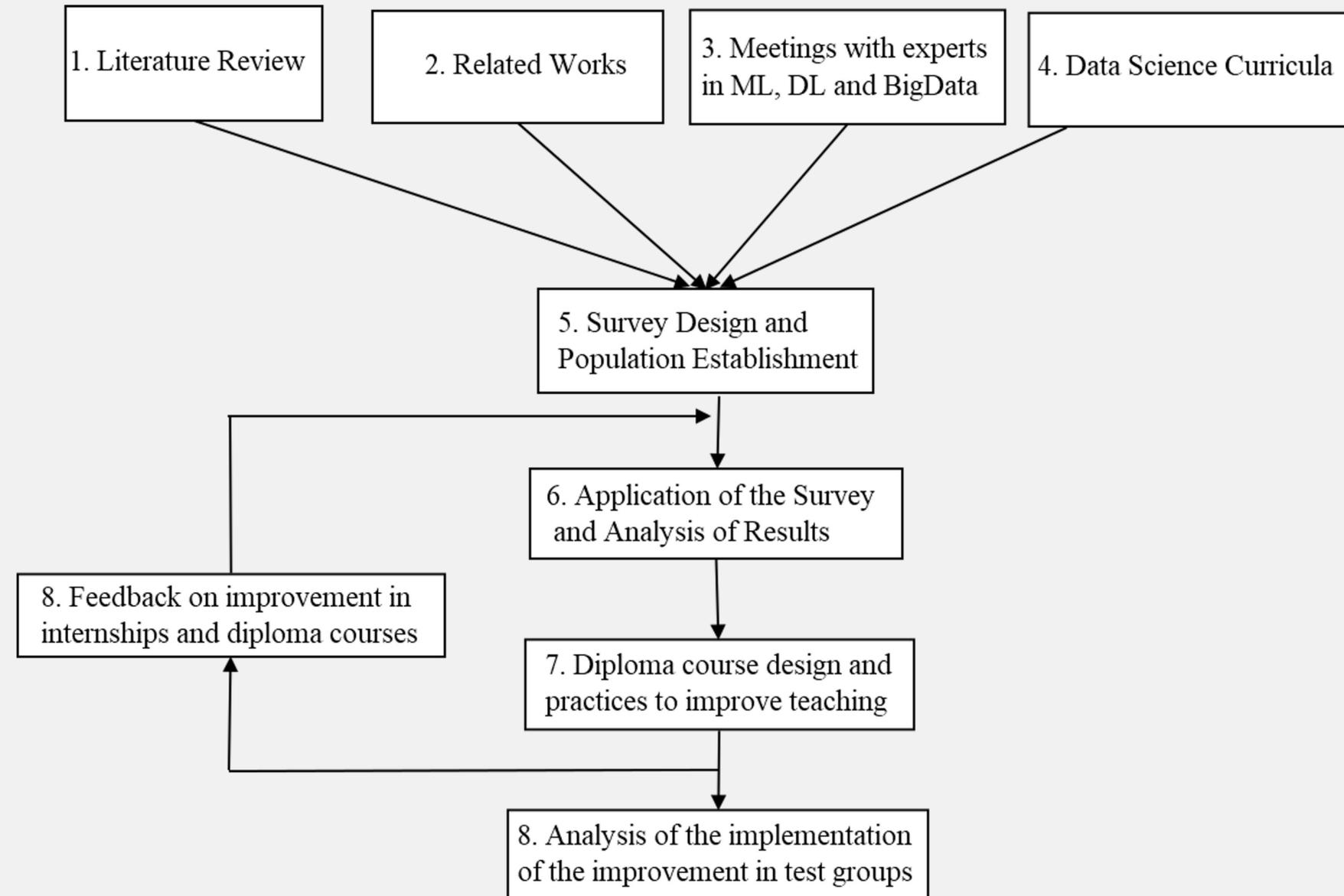
Send the survey to qualified students and teachers.



Collect the data and analyze them.

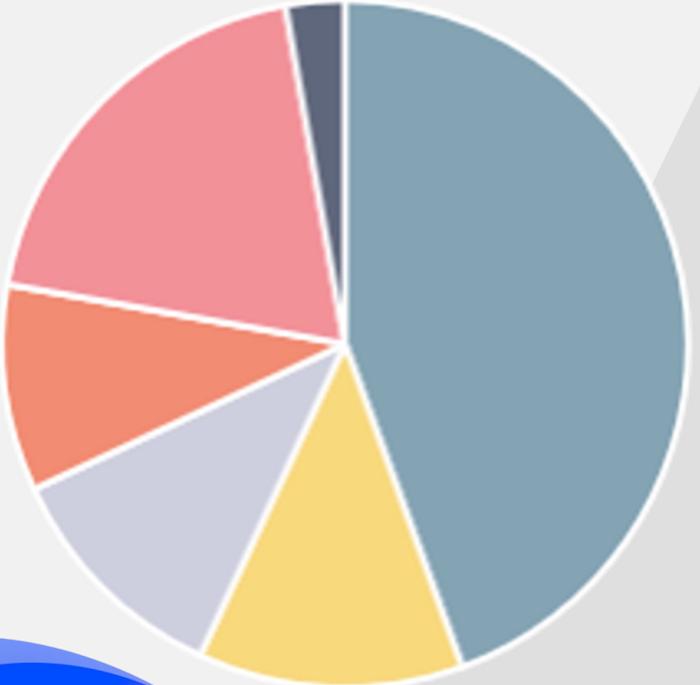
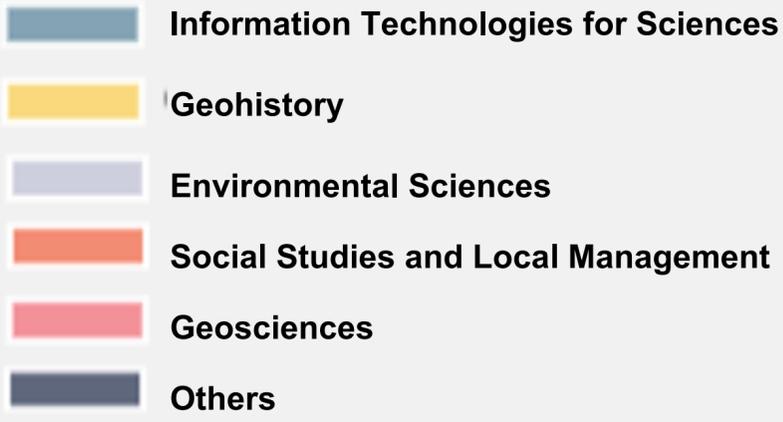


Select the appropriate topics to implement practices in Python.



Methodology

Developing the Survey



The base population size of 600 students and teachers from careers related to ML, DL, and Big Data

Online survey to compile and process all submissions



Random students and academic staff were selected across the UNAM campus in Morelia

Factors to Improve Teaching

Diploma course

MODULE 1- Machine Learning (ML).
"Theory and Practice for the Improvement
of the Teaching of ML Applied to Data
Science"

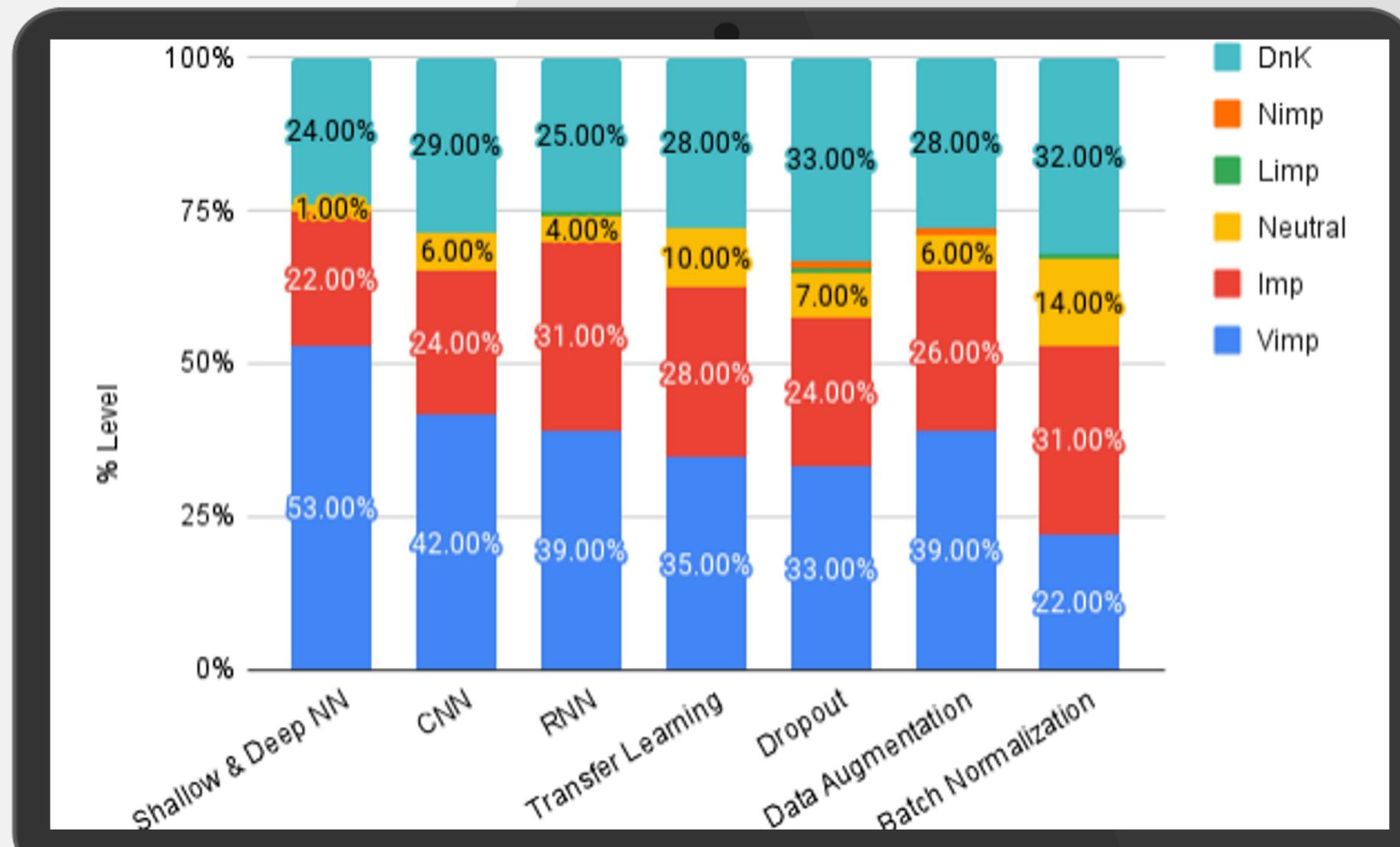
MODULE 2- Deep Learning (DL). "Theory
and Practice for the Improvement of the
Teaching of DL Applied to Data Science"



Survey Results

Quantifying and analyzing results

Based on the survey, we identified 4 important axes for classifying topics to cover, and based on the skill levels shown by the students and teachers, we began to design the first practices to fit the average knowledge of the respondents



DL axis and related topics

Designing the ML Module

10

Practices that were planned and designed to cover ML needs for all students and teachers

45

Hours in an intensive inter-semester diploma course

#	Name	Dataset	Evaluation Metric
1	Classification using decision trees	Titanic passengers	Accuracy and/or Fbeta Metrics
2	Housing cost prediction	California Housing	RMSE and/or MAE
3	k-Nearest Neighbors	Water wells	Precision Score Fbeta
4	k-Means	Online Retail k-means & Hierarchical Clustering	Not applicable
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Designing the DL Module



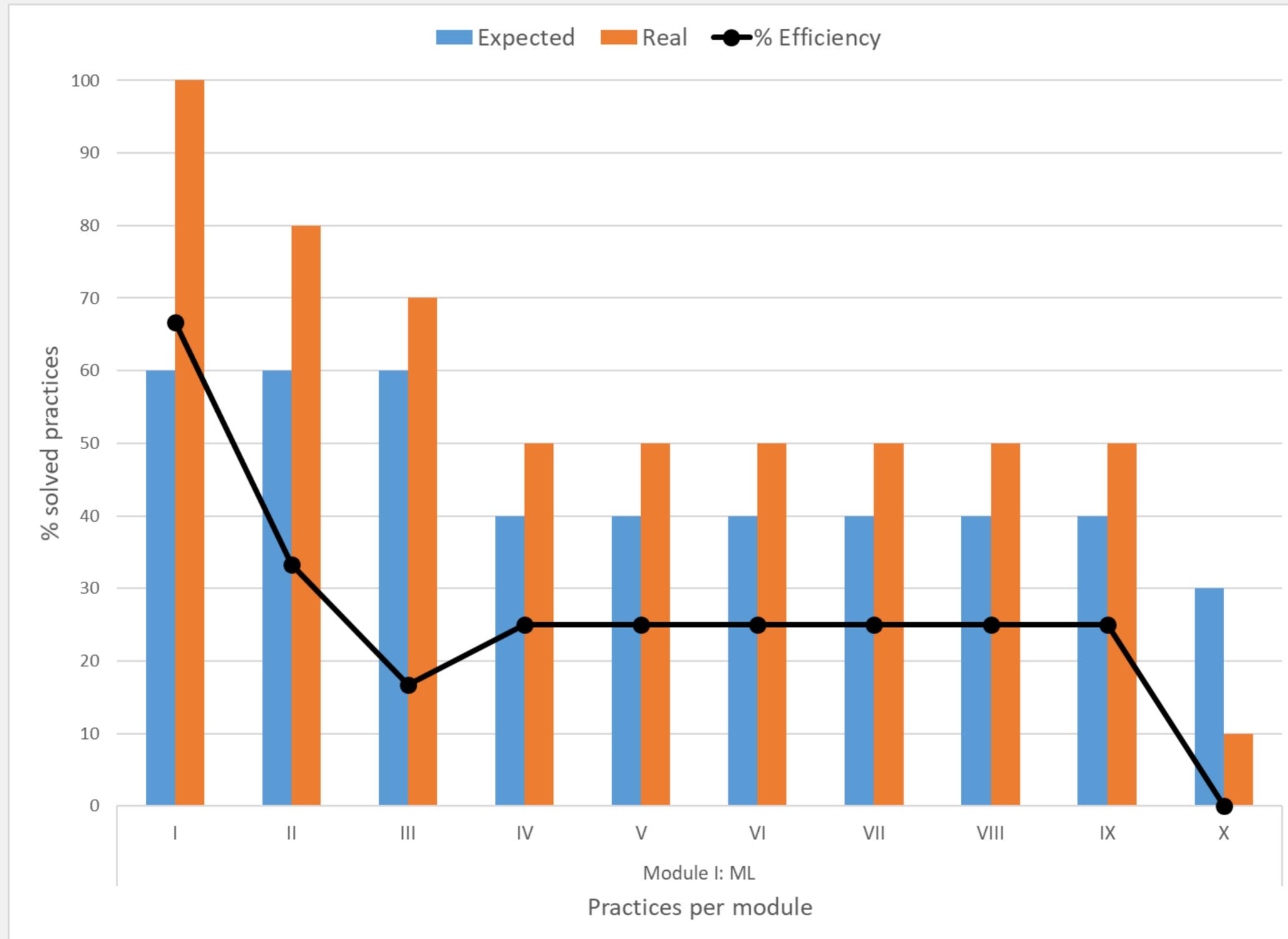
Practices that were planned and designed to cover DL needs for all students and teachers



Hours in an intensive inter-semester diploma course.

#	Name	Dataset	Evaluation Metric
1	Binary classification with CNN	800 images of mosquitoes, UNAM	Accuracy
2	Binary classification with CNN	Covid-19 Pneumonia Screening	Precision and Confusion Matrix
3	CNN & Data Augmentation	Ship Classification	Precision and Confusion Matrix
4	RNN	Sarcasm Detection	Accuracy and Precision
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ML Teaching Effectiveness



Results and Conclusions



ACKNOWLEDGMENT



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