Migration to Microservices: A Comparative Study of Decomposition Strategies and Analysis Metrics



June 26th-30th Nice, France



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Outline:

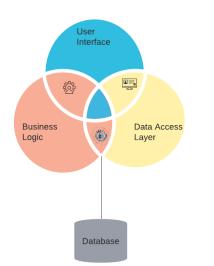
- Context
- Related Work
- Proposed Approach
- Research Questions
- Conclusion





Context 1/2

Monolith architecture



Microservice architecture



🗙 Lack of Scalability

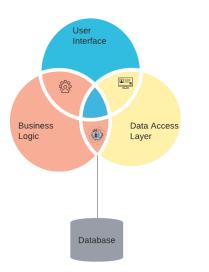


X Limited Development Team Autonomy

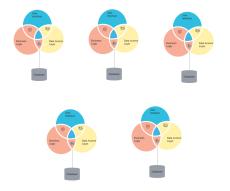


Context 1/2

Monolith architecture



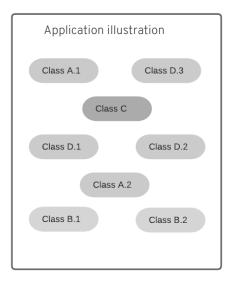
Microservice architecture

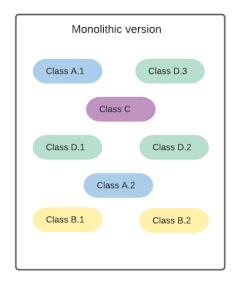


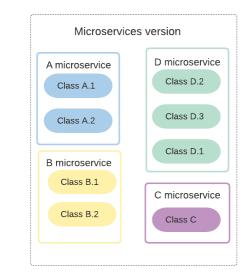
Scalability and flexibility
 Fault Isolation and Resilience
 Team Autonomy



Context 2/2









Decomposing process

Monolithic application

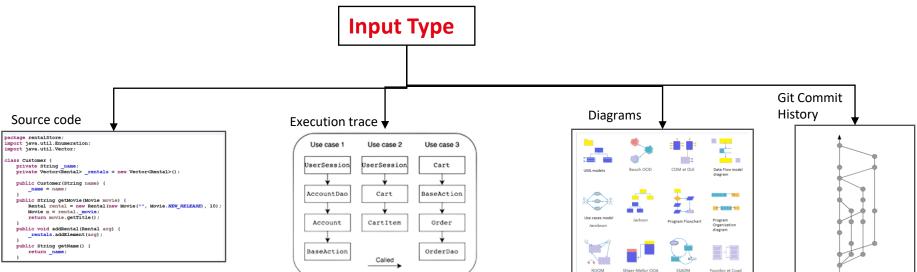




Microservice application



Related Work 1/2



Bunch B. S. Mitchell and S. Mancoridis, "On the evaluation of the bunch search-based software modularization algorithm," Soft Computing 2008

HierDecomp K. Sellami, M. A. Saied, and A. Ouni, "A hierarchical dbscan method for extracting microservices from monolithic applications," in The International Conference on Evaluation and Assessment in Software Engineering 2022 **Mono2Micro** K. Kalia, X. Jin, K. Rahul, S. Saurabh, V. Maja, and B. Debasish, "Mono2micro: A practical and effective tool for decomposing monolithic java applications to microservices," Proceedings of the 29th ACM Joint Meeting on European Software Engineering Conference and Symposium on the Foundations of Software Engineering, 2021.

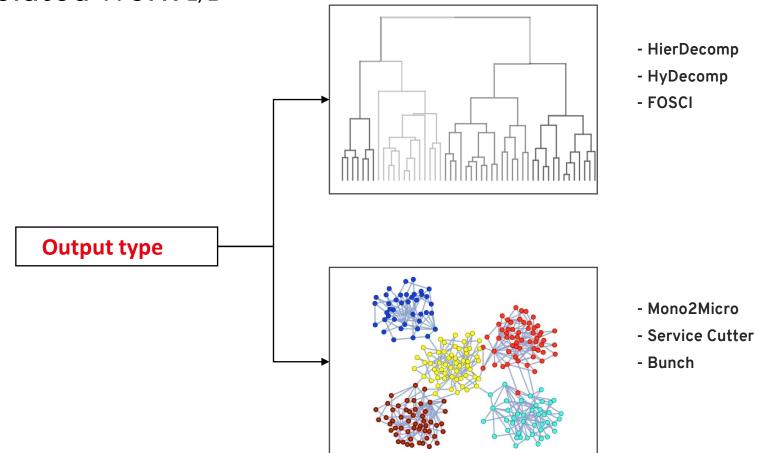
FOSCI W. Jin, T. Liu, Y. Cai, R. Kazman, R. Mo, and Q. Zheng, "Service candidate identification from monolithic systems based on execution traces," IEEE Transactions on Software Engineering2021

Service cutter

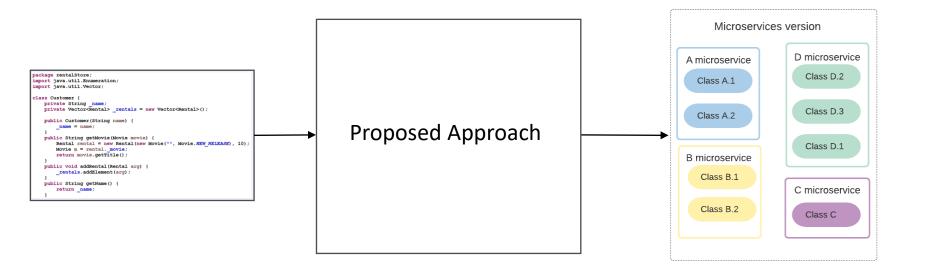
Michael Gysel, Lukas Kolbener, Wolfgang Giershe, Olaf Zimmermann, "Service Cutter: A Systematic Approach to Service Decomposition" LNPSE,2016 **MEM** G. Mazlami, J. Cito, and P. Leitner, "Extraction of microservices from monolithic software architectures," in 2017 IEEE International Conference on Web Services (ICWS)



Related Work 2/2







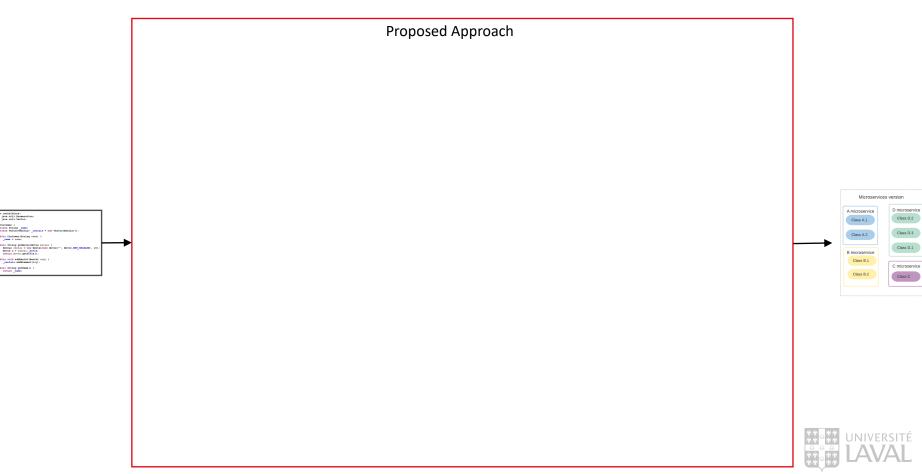


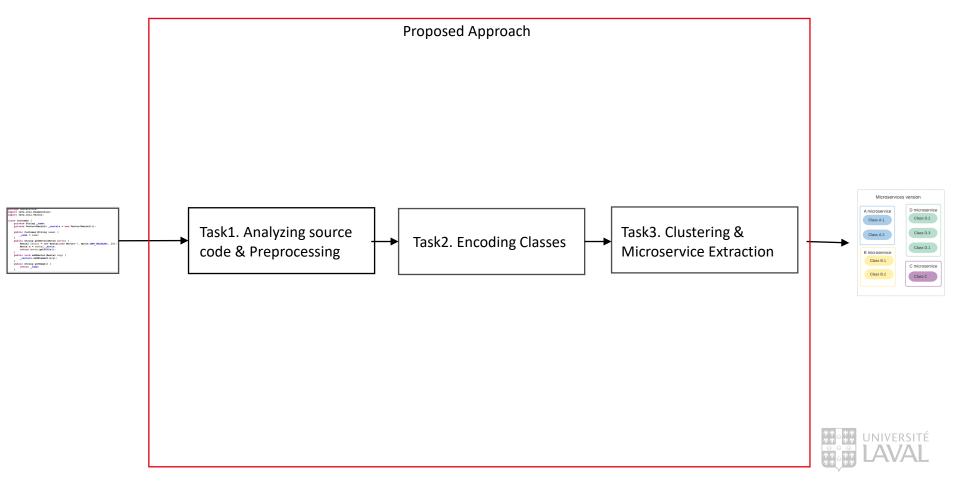
se Customer (private String _______ private VectorOmental> __rentals = new VectorOmental>().

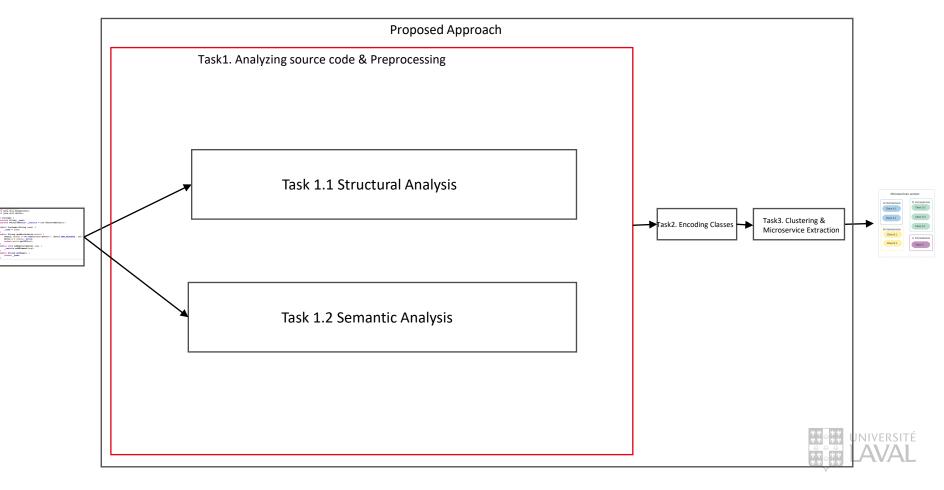
public Customer(String name) (

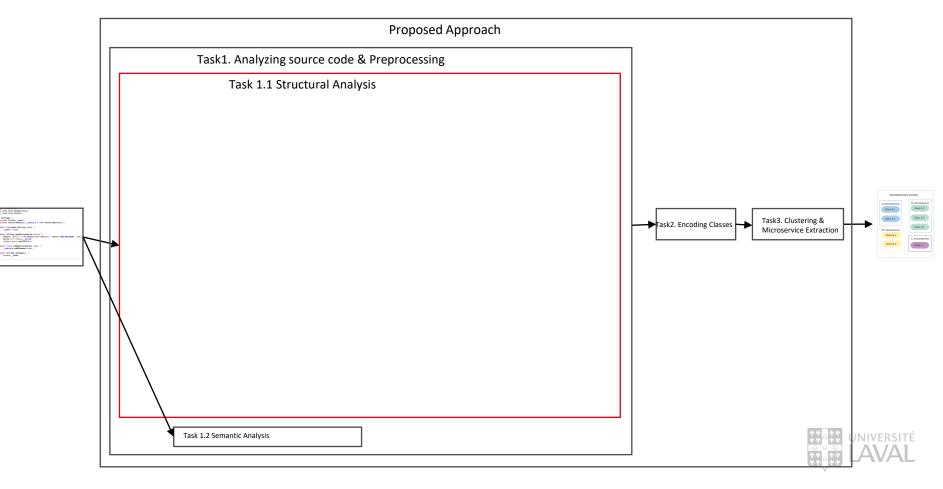
blic void addRestal(Restal arg) __ventals.addRiesest(arg)/

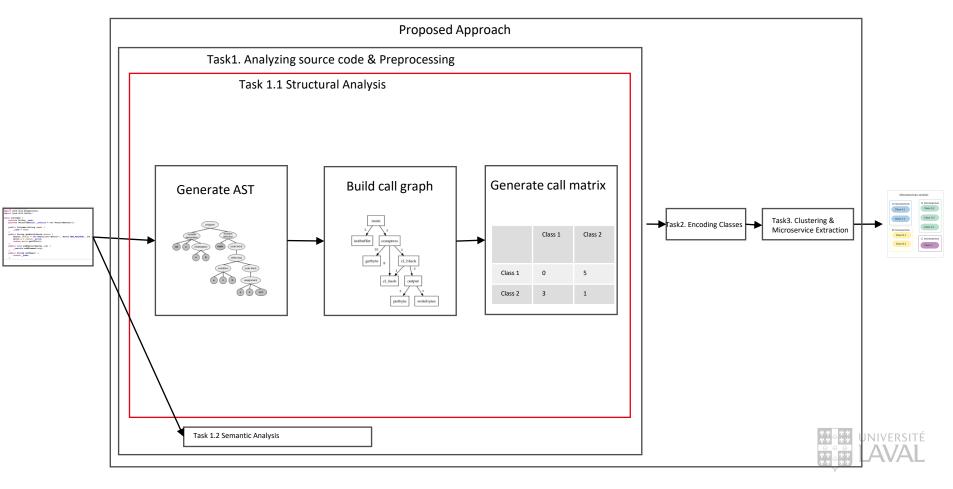
) public String getNume() (return _____)

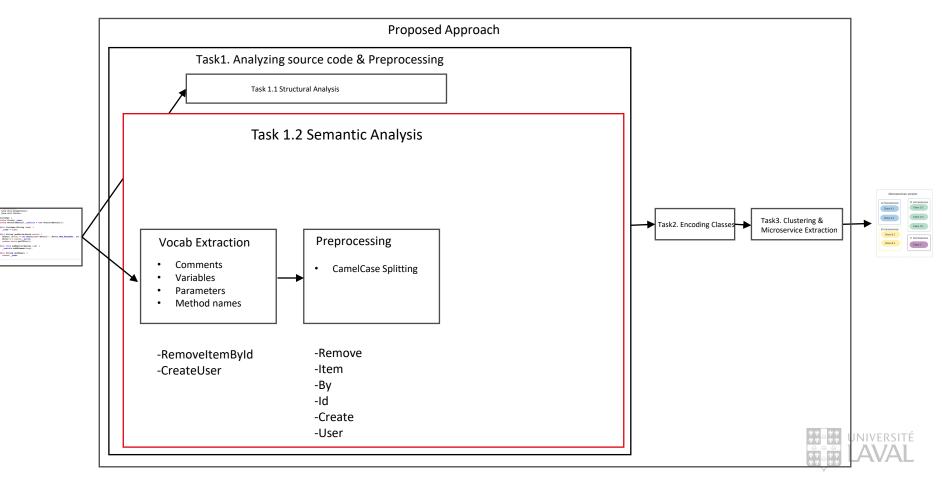


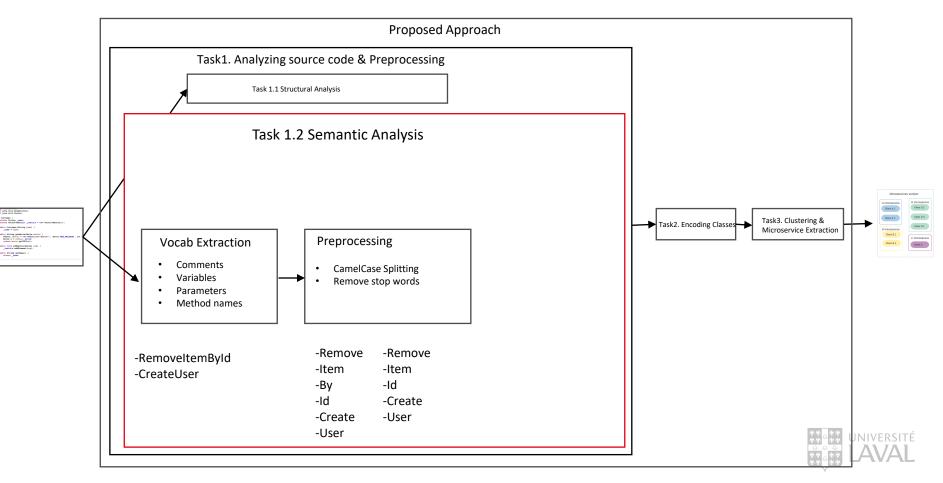




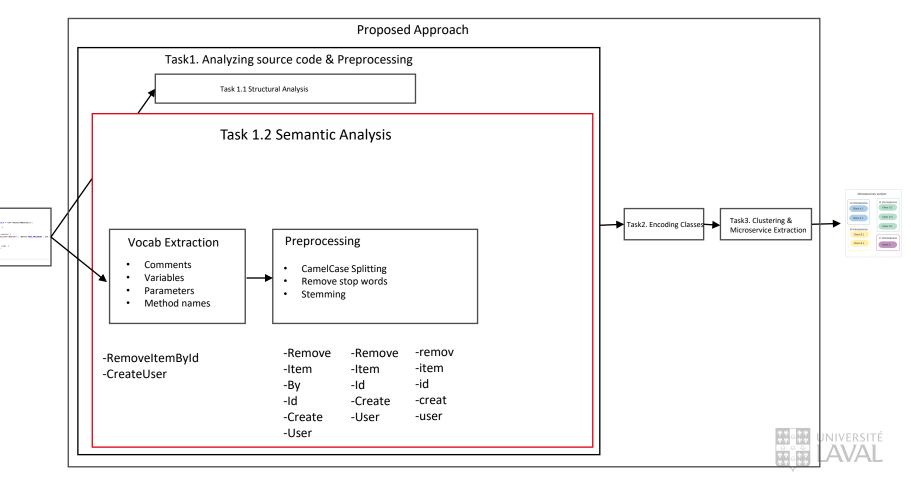


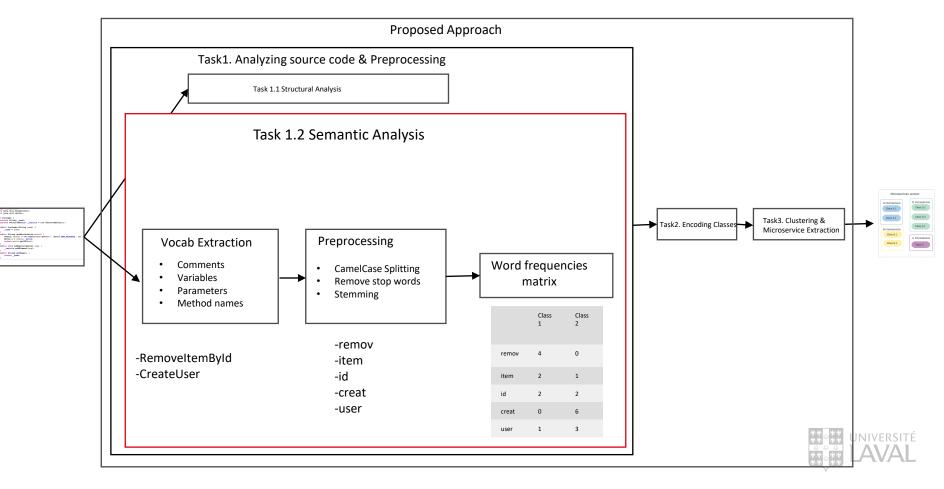




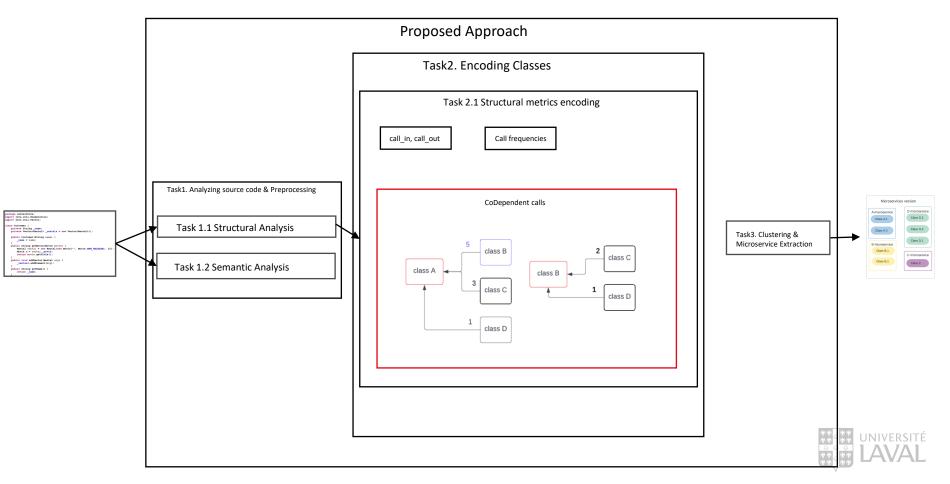


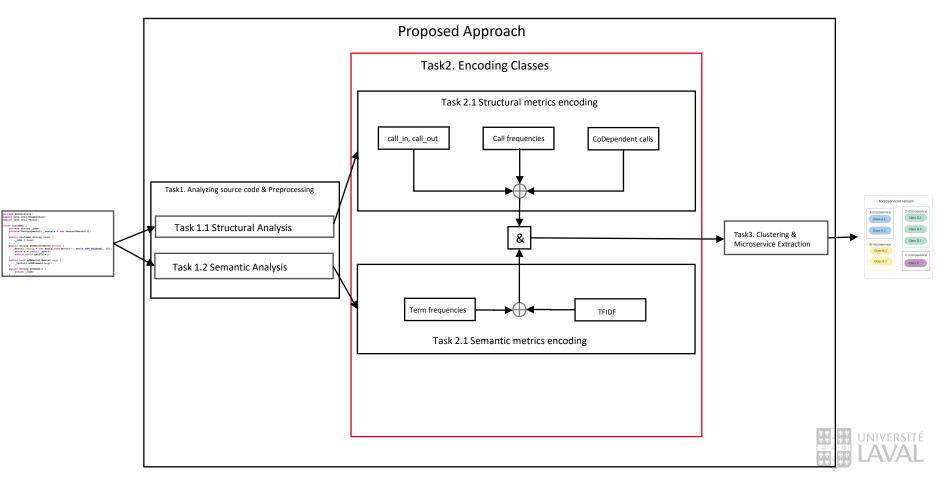
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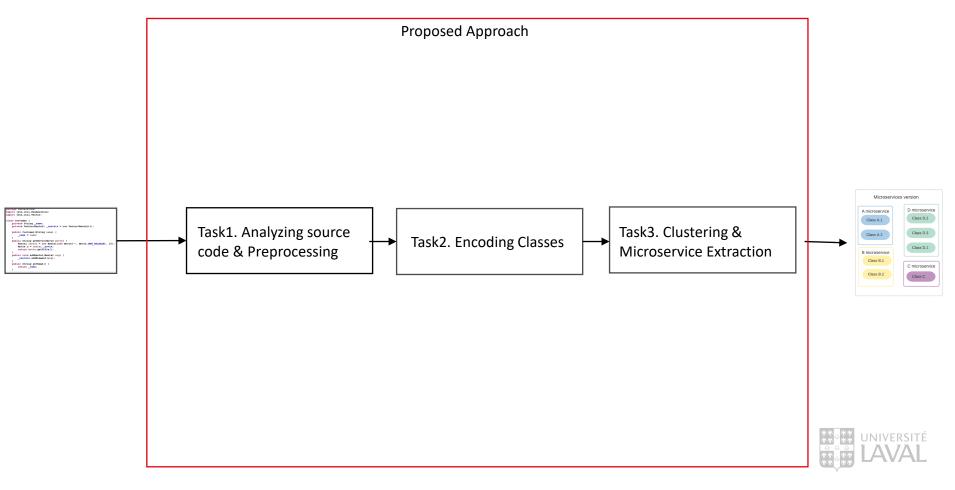


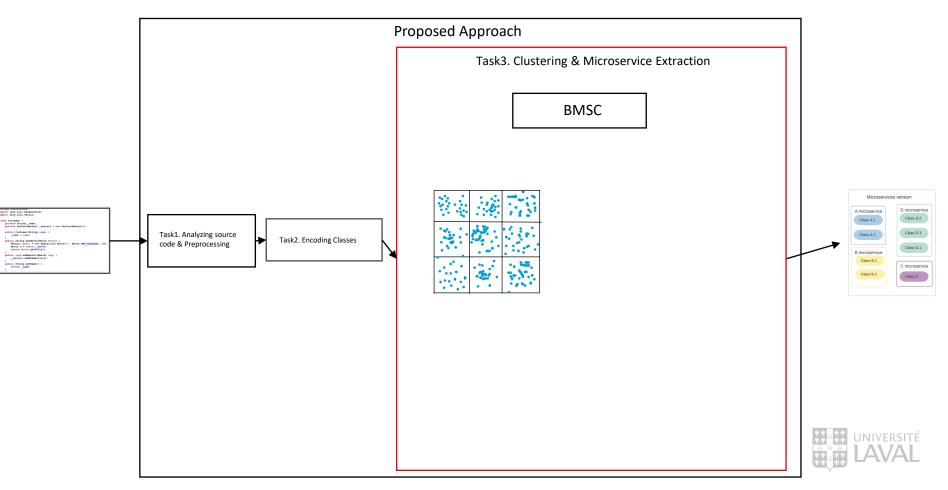


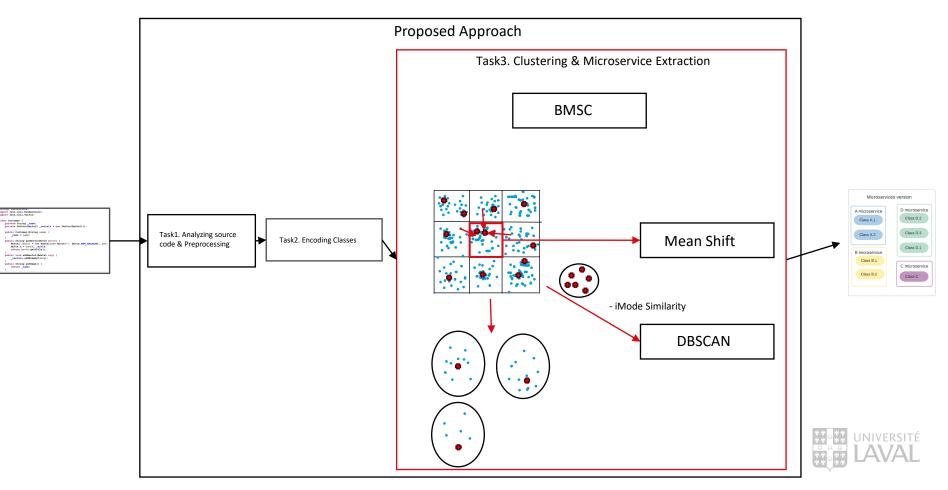
		Proposed Approach		
		Task2. Encoding Classes		
		Task 2.1 Structural metrics encoding call_in, call_out Call frequencies CoDependent calls		
<pre>provide the second second</pre>	Task1. Analyzing source code & Preprocessing Task 1.1 Structural Analysis Task 1.2 Semantic Analysis	CoDependent calls	Task3. Clustering & Microservice Extraction	Microsovices version A incoursive (and a) B incoursive (and b) (and b) (and b) (and b) (and b) (and b) (and b)
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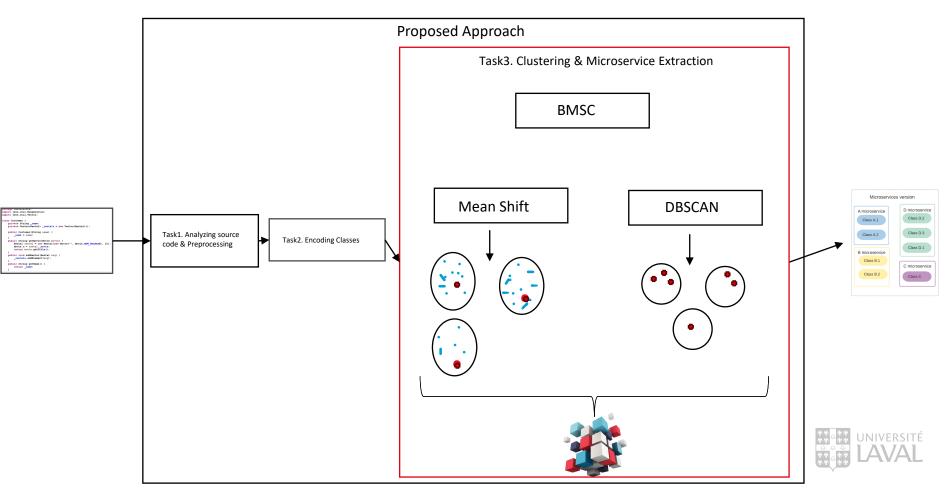


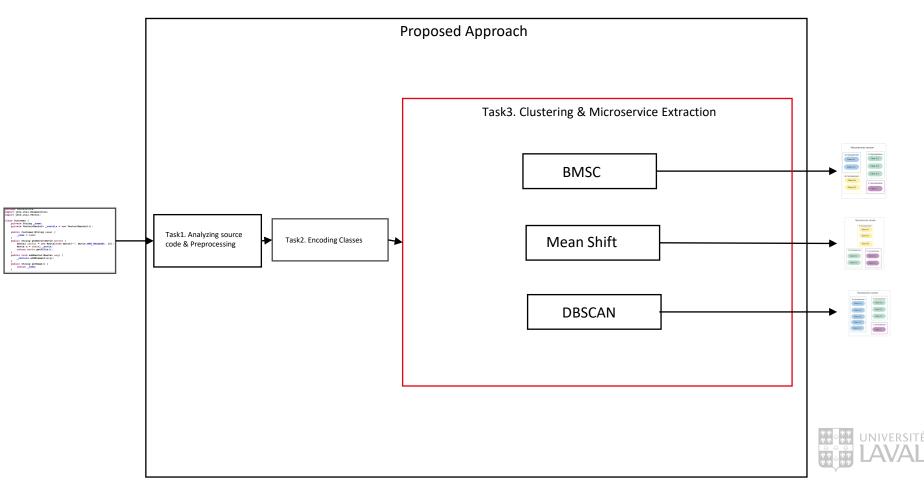












What is the most effective and promising option among the various choices in our approach?

Project:

DayTrader: 118 classes

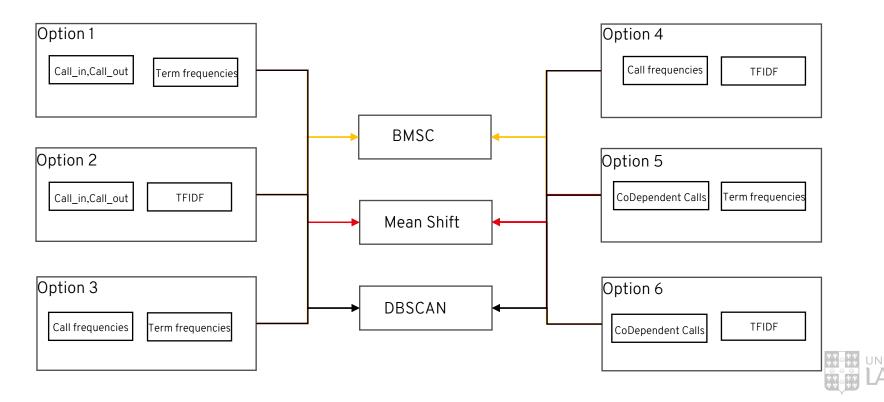




How can we evaluate the performance of the strategies?: Metrics

- ↑ Inter Call Percentage (ICP)
- ↑ InterFace Number (IFN)
- ↑ Non-Extreme Distribution (NED)





Evaluation Results using Mean Shift Algorithm

		-	-		-	
Metrics	Option 1	Option 2	Option 3	Option 4	Option 5	Option 6
SM	0.8526	0.7853	0.7944	0.8614	0.8575	0.8742
IFN	1.235	1.8	1.277	1.0454	1.0	1.214
ICP	1.0	0.9	1.0	1.0	1.0	1.0
NED	1.0	0.9	1.0	1.0	1.0	1.0
# microservices	17	10	18	22	21	14
size of the largest micro	98	102	97	92	97	104

Evaluation Results using DBSCAN Algorithm

Metrics	Option 1	Option 2	Option 3	Option 4	Option 5	Option 6
SM	0.120	0.1085	0.2702	0.2718	0.2487	0.1116
IFN	0.120	0.1085	0.2702	0.2718	0.2487	0.1116
ICP	0.3244	0.1426	0.1591	0.2859	0.3482	0.0079
NED	0.5	0.666	1.0	0.5	0.5	0.333
# microservices	2	3	2	2	2	3
size of the largest micro	108	86	116	113	113	104

Evaluation Results using BMSC Algorithm

Metrics	Option 1	Option 2	Option 3	Option 4	Option 5	Option 6
SM	0.3696	0.3435	0.3887	0.4697	0.40545	0.4052
IFN	1.0344	1.250	1.0370	0.9677	1.0769	1.318
ICP	0.6500	0.591	0.618	0.6432	0.6257	0.639
NED	0.7241	0.6666	0.7037	0.7419	0.6538	0.636
# microservices	29	24	27	31	26	22
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Option 6 is the optimal approach for all three clustering algorithms. It uses co-dependent calls metric as structural information and TF-IDF vector as semantic information. Consequently, our work will continue to focus on this strategy.

Evaluation Results using Mean Shift Algorithm

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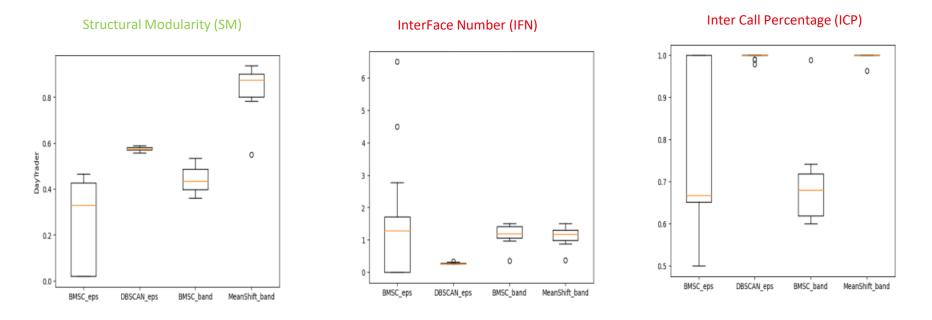
How does the stability and robustness of BMSC algorithms compare to that of Mean Shift and DBSCAN?

Hyperparams:

- Bandwidth
- Epsilon
- MinPts =5



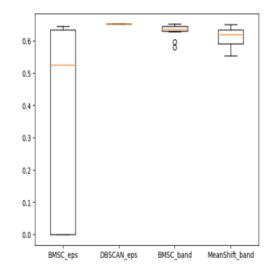
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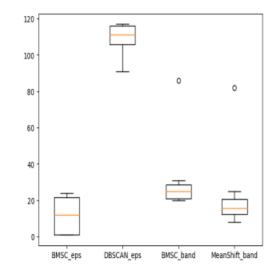


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How does the stability and robustness of BMSC algorithms compare to that of Mean Shift and DBSCAN?: Results

Non Extreme Distribution(NED)

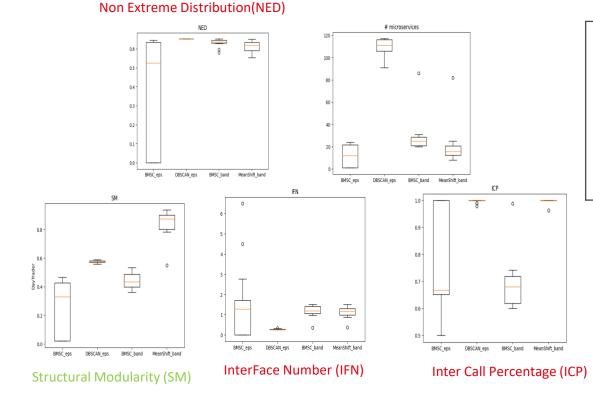




microservices



How does the stability and robustness of BMSC algorithms compare to that of Mean Shift and DBSCAN?: Results



In contrast to the literature, our analysis suggests that for our case, BMSC is more susceptible to the selection of its hyperparameters, specifically the epsilon parameter, compared to DBSCAN and Mean Shift when used independently. However, BMSC demonstrates greater consistency in the resulting decompositions across hyperparameter variations.



Conclusion

