SASS: Self-Adaptation using Stochastic Search

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Smart Grids



Smart Buildings



Smart Traffic Systems



Smart Farms

Stochastic Search

Deterministic

Stochastic





Stochastic Search



Stochastic Search

- •B. H. C. Cheng, A. J. Ramirez, and P. K. McKinley, "Harnessing evolutionary computation to enable dynamically adaptive systems to manage uncertainty"
- G. G. Pascual, M. Pinto, and L. Fuentes, "Run-time adaptation of mobile applications using genetic algorithms"
- •P. Zoghi, M. Shtern, and M. Litoiu, "Designing search based adaptive systems: A quantitative approach"
- •M. Harman, Y. Jia, W. B. Langdon, J. Petke, I. H. Moghadam, S. Yoo, and F. Wu, "Genetic improvement for adaptive software engineering (keynote)"

Self-Adaptive System



Requirements for Stochastic Techniques

Possible Solution Creation

Randomly create a tree of tactics

- Objective Function
 - Example: -20 responseTime $_{norm}$ cost $_{norm}$ + quality $_{norm}$ time $_{norm}$
 - Use PRISM to evaluate the score of each final state in the plan and the probability of reaching each final state



ZNN.com



Proof of Concept

Basic Action	Precondition	Cost (\$)	Resp. Time	Time	Failure Rate
Add S_1 Server	$S_1 < S_1^{max}$	15	-5	600	0.1
Add S_2 Server	$S_2 < S_2^{max}$	20	-5	600	0.1
Remove S_1 Server	$S_1 > 1^{-1}$	-15	5	600	0.1
Remove S_2 Server	$S_2 > 1$	-20	5	600	0.1
Add DB_A Thread	$T_A < T_A^{max}$	0	-2	180	0.2
Add 2 DB_B Threads	$T_B < T_B^{max}$ -1	0	-1	180	0.2
Increase Quality	Quality set low	0	2S	1	0.3
Decrease Quality	Quality set high	0	-2S	60	0.3

Plans

[(if (add-server(1))
 (add-database-thread(A,1))
 (decrease-quality))),
add-database-thread(B,2)]



Experiments

- Improving a bad plan
 - Show that subpar inputs do not cause subpar results
- Comparing different utility functions
 - Understand how changes in the utility function affects plans
- Planning with similar utility functions
 - Can plans be used to create new plans for a similar goal?

Improving a Bad Plan

- Utility Function: -20 responseTime_{norm}
 - $-\cos t_{norm} + quality_{norm} time_{norm}$
- Initial Plan: $-S_2$ Server + 1 DB₄ + S₁ Threads Server × × - S₁ + 2 DB_B - S₁ + 2 DB_B Server Server Threads; Threads: + 2 DB_R + 2 DB_B x × Threads Threads + 2 DB_B +1 DB_△ + 2 DB_R +1 DB_△ Threads Thread Threads Thread

Improving a Bad Plan

• Results:



Comparing Utility Changes

- Utility Function : -10 responseTime_{norm}
 - $\text{cost}_{\text{norm}} + \text{quality}_{\text{norm}} \text{time}_{\text{norm}}$
- Results:



Comparing Utility Changes

- Utility Function : -2 responseTime_{norm}
 - $-2 \text{ cost}_{norm} + \text{ quality}_{norm} \text{ time}_{norm}$
- Results:



Planning with Similar Utility Functions

- Utility Function: -10 responseTime_{norm}
 - $\text{cost}_{\text{norm}} + \text{quality}_{\text{norm}} \text{time}_{\text{norm}}$



Experiment Conclusions

- The planner can handle erroneous user provided plans
- The planner can handle multiple objectives
- The planner can provide unexpected knowledge about the search space
- The planner can use information from previously generated plans to make new plans

Future Ideas

- Test the planner on a system with more tactics
 Compare to deterministic planners
- Incorporate feedback from the system monitor
 - If adaption fails, is it likely to fail again?
 - Partially effective adaptations timing issues
- Adapting similar plans to a new situation
- Catalogue when stochastic techniques are effective
- Improve human trust in plans/ stochastically generated plans

Summary

- Stochastic search shows promise for handling the future complexity of self-adaptive systems
- Demonstrate the benefits of stochastic search with a proof of concept genetic programming planner
- 3 experiments demonstrate the potential of our planner
- There are many research problems for applying stochastic search to self adaptive systems