

## Research in Experimental Computer Science

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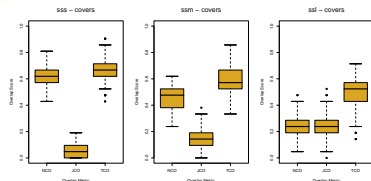
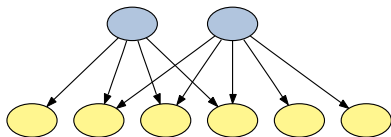
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Allegheny College summeR reSearch Series (ACRoSS), July 2008  
Featuring images from [www.campusbicycle.com](http://www.campusbicycle.com) and [www.pdclipart.org](http://www.pdclipart.org)

# What is Experimental Computer Science?

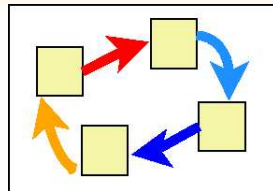
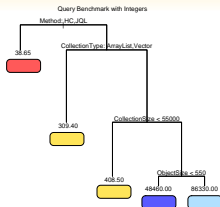


Solve Problems with Algorithms

Detailed Empirical Results

**Implement** and empirically **evaluate** the efficiency and effectiveness of **algorithms** that solve real-world **problems**

# What is Experimental Computer Science?



Statistical Analysis Techniques

Working Computational Artifacts

After analyzing **gigabytes** of data, **publish** results and **release** software **tools** that are useful to **academics** and **industrialists**

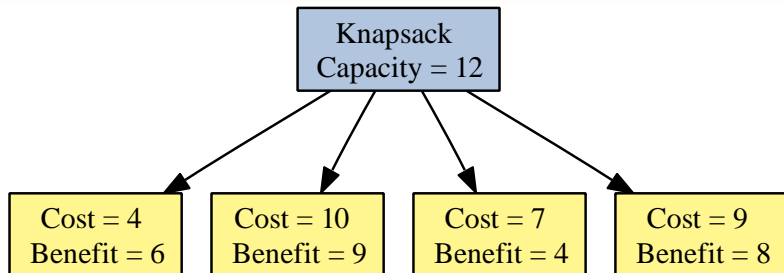
**Computer software as community service**

# Backpacks and Bicycles



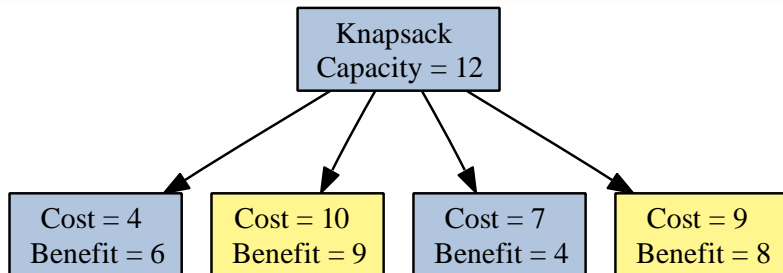
**Goal:** Find a backpack that will support your commute to work or school

# 0/1 Knapsack Problem



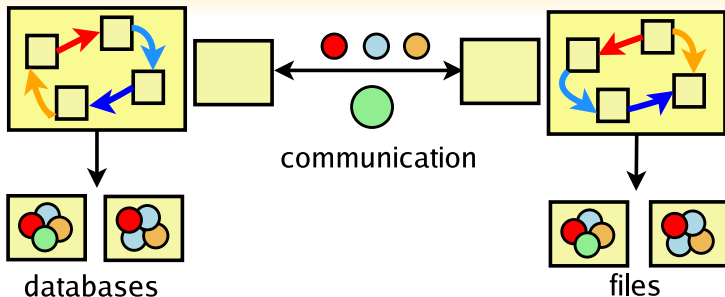
- **Question:** Can you select items so that you **maximize** the benefit while ensuring that the cost does not **exceed** the capacity?
- This problem is **NP-complete** (see Garey and Johnson) and yet it also has many practical applications in both **software** and **finance**

# 0/1 Knapsack Problem



- **Question:** Can you select items so that you **maximize** the benefit while ensuring that the cost does not **exceed** the capacity?
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# Complexity and the Software Crisis



*"Software entities are more complex for their size than perhaps any other human construct"*

- **Frederick Brooks** (Professor of Computer Science at the University of North Carolina – Chapel Hill)

# Regression Testing Techniques

Before



After



Reduction Prunes the Test Suite

Before



After



Prioritization Reorders the Tests

It is **expensive** to run a test suite  $T = \langle T_1, \dots, T_n \rangle$ . **Prioritization** searches through the  $n! = n \times n - 1 \times \dots \times 1$  orderings for those that **maximize** an objective function like **coverage** or **fault detection**.



# Prioritizing When Memory is Constrained



Frequent Memory Rewrites



High Testing Costs

Frequent **reads** and **writes** to memory may **increase** execution time by as much as **600%** when a Java application executes on a virtual machine with a **small heap**.

**Solution:** maximize memory **reuse** between test cases

# The Impact of Test Ordering

	$m_1$ 30	$m_2$ 30	$m_3$ 30	$m_4$ 30	$m_5$ 30	$m_6$ 30	Test Size
$T_1$	•	•	•				90
$T_2$				•	•	•	90
$T_3$	•	•	•				90
$T_4$				•	•	•	90
$T_5$	•	•					60

- $T = \langle T_1, T_2, T_3, T_4, T_5 \rangle$  transfers **750** units to and from memory
- $T' = \langle T_2, T_4, T_1, T_3, T_5 \rangle$  only loads and unloads **180** units

# Challenges of Test Prioritization

## Real-World Suites Have Hundreds of Tests

- $n!$  possible solutions
- For 100 test cases,  $9.33262154 \times 10^{157}$  possible solutions
- It takes 2.22 seconds to evaluate a solution permutation
- Finding the answer would take  $6.52846694 \times 10^{149}$  years

## Problem Formulation

- **Intelligently** comb the search space for an **effective** ordering
- Can **efficient** techniques identify **good** prioritizations?
- Will the prioritizers work properly in **real-world** software development environments?

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# Cheapest Way to Travel the World?



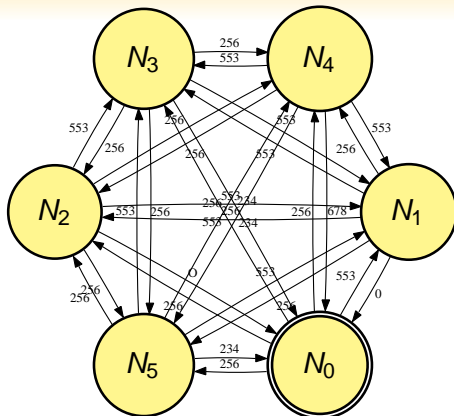
Cheapest way to see Seattle, Rio, Shanghai, Mumbai, and Sydney.

# Cheapest Way to Travel the World?



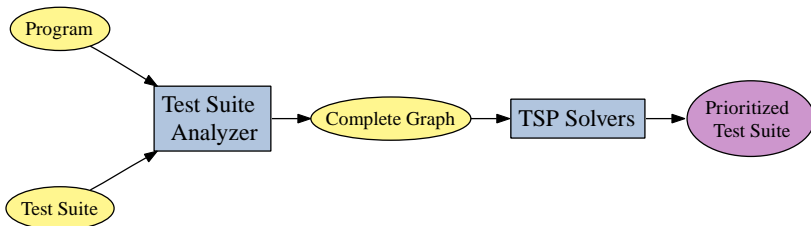
$5! = 120$  possible solutions

## Classic Problem in Graph Theory



**Goal:** Find the least weight Hamiltonian path through a complete graph

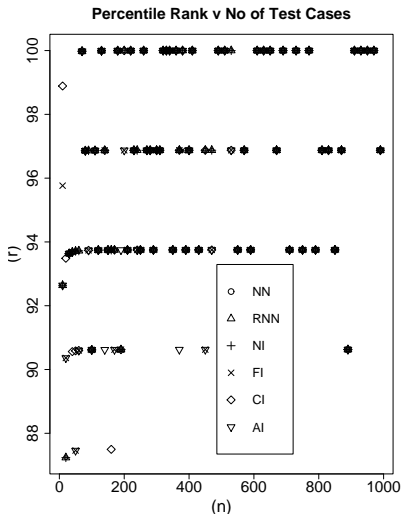
# Prioritizing with Hamiltonian Paths



- Is it possible to **efficiently** create test prioritizations by **solving** the “find the cheapest way to travel the world” problem?
- If yes, then are the orderings **effective**?



# How Good are TSP Solvers?



- Implemented by researchers at the University of Vienna
- Six key algorithmic techniques
- All algorithms produce 80+ percentile rankings for the path cost metric
- Most expensive algorithm takes 200s for  $n = 1000$

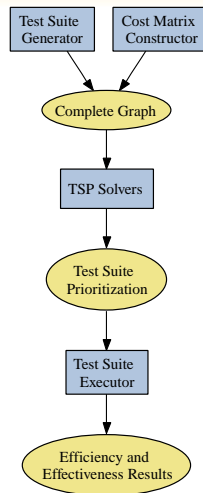
# Experiment Design

## ● Project Metrics:

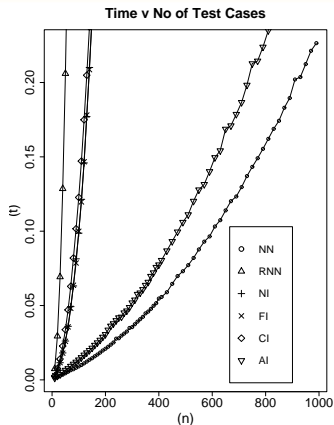
- Lines of Code: 5,274
- Data Files: 15,180
- Data Files Size: 4.5 GB

## ● Key Implementations:

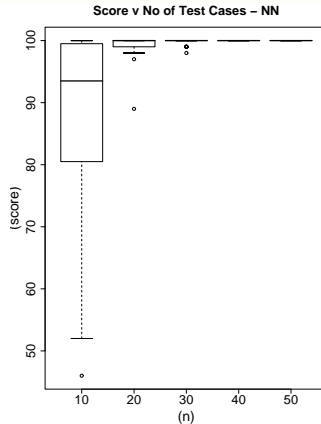
- Synthetic Test Suite Generator
- Cost Matrix Constructor
- Test Suite Executor



# Empirical Results



Efficient Prioritizers



High Percentile Rankings

# What is a Genetic Algorithm?

A **search-based** approach to test suite prioritization

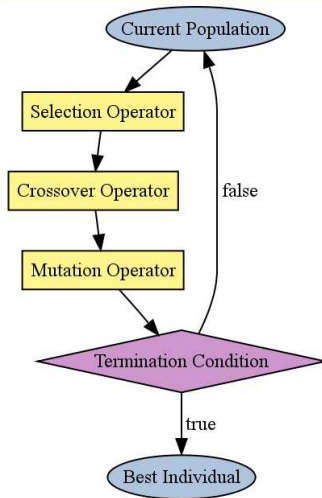
Parts of a Genetic Algorithm:

- **Data Structures:**

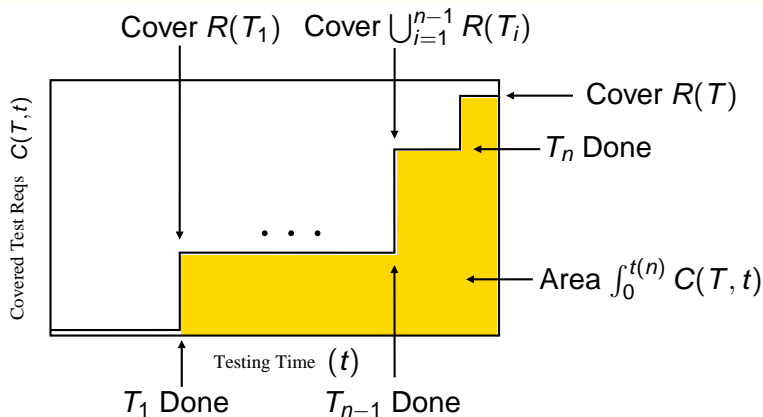
- Chromosome
- Individual
- Population

- **Functions:**

- Selection Operator
- Crossover Operator
- Mutation Operator
- Termination Condition
- Fitness Function



# What is Coverage Effectiveness?



- Prioritize to **increase** the CE of a test suite  $CE = \frac{\text{Actual}}{\text{Ideal}} \in [0, 1]$

# Genetic Test Suite Prioritizer

- **Motivation:** There are known instances where greedy techniques always yield sub-optimal orderings. Few experiments have studied the efficiency and effectiveness of search-based techniques.
- **Goals:** Identify configurations of the genetic algorithm that produce desirable results. Outperform random search.

## Project Statistics:

- 6,369 lines of code
- 6 mutation operators
- 7 crossover operators
- 3 selection operators
- 3 fitness transformation operators

# Experiment Design

## ● Metrics:

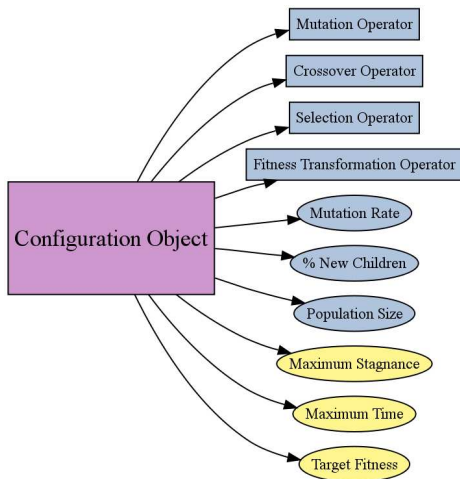
- Runtime of prioritization technique
- Coverage effectiveness of final test ordering

## ● Data Sets:

- 9 real-world
- 54 synthetic

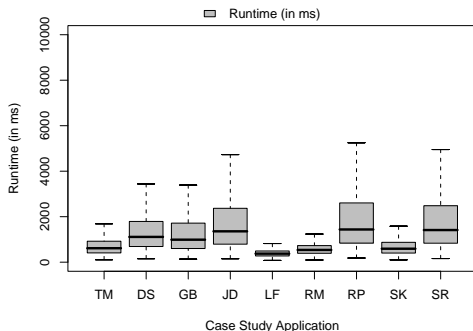
## ● Configurations:

- 10,206 configurations
- 91,854 real-world experiments
- 551,124 synthetic experiments

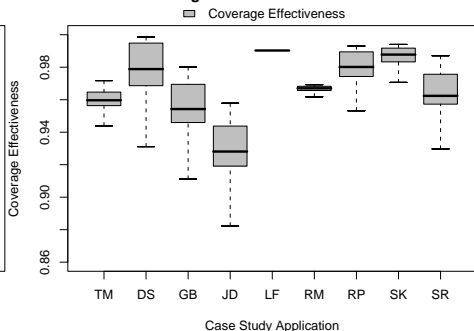


# Empirical Results

## Runtime of Search-Based Prioritizer



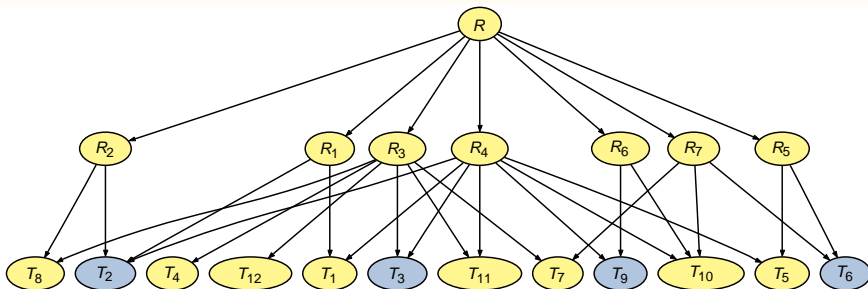
## Coverage Effectiveness of Test Suite



Preliminary investigation indicates that the genetic algorithm can produce **better** results in **less time** than random search

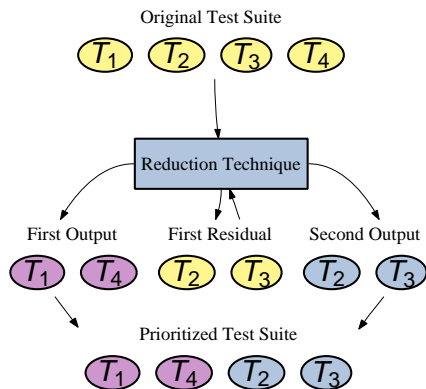


# Finding the Overlap in Coverage



- $R_j \rightarrow T_i$  means that requirement  $R_j$  is **covered by** test  $T_i$
- $T = \langle T_2, T_3, T_6, T_9 \rangle$  covers **all** of the test requirements
- Test suite **reduction** discards the test cases that **redundantly** cover the test requirements

# Incorporating the Costs of a Test Case



- Harrold, Gupta, Soffa (HGS)
- Delayed Greedy (DGR)
- Traditional Greedy (GRD)
- 2-Optimal Greedy (2OPT)

**Hypothesis:** Using the execution **time** of a test case can **improve** the reduced and prioritized test suites

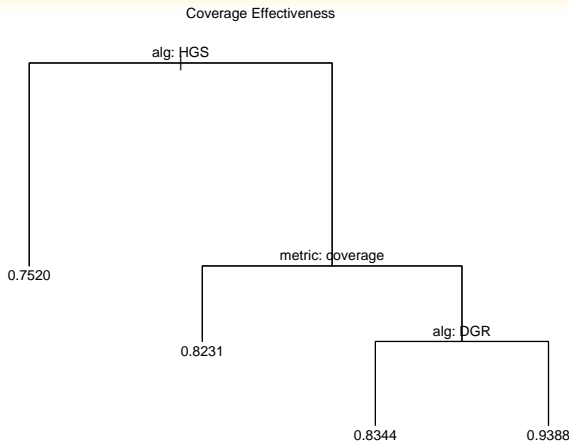
Compare (i) **greedy choices** (cost, coverage, and ratio) and (ii) **algorithms**

# Greedy Choices Impact Effectiveness

	$R_1$	$R_2$	$R_3$	$R_4$	$R_5$	Execution Time
$T_1$	✓	✓	✓	✓		4
$T_2$			✓	✓		1
$T_3$		✓				1
$T_4$	✓				✓	1

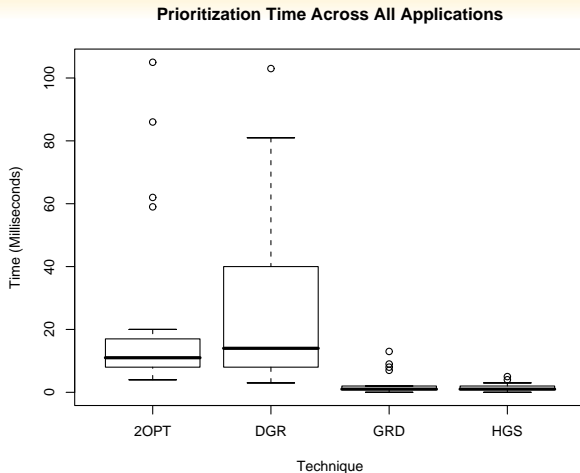
Greedy-by	$T_r$	$time(T_r)$	$T_p$	CE
coverage	$\langle T_1, T_4 \rangle$	5	$\langle T_1, T_4, T_2, T_3 \rangle$	0.400
time	$\langle T_2, T_3, T_4 \rangle$	3	$\langle T_2, T_3, T_4, T_1 \rangle$	0.714
ratio	$\langle T_2, T_4, T_3 \rangle$	3	$\langle T_2, T_4, T_3, T_1 \rangle$	0.743

# Empirical Results: Effectiveness



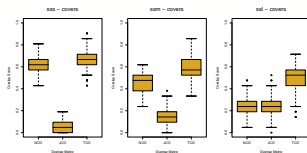
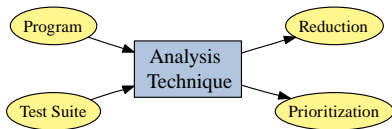
Using **ratio** and **time** improves the CE of the prioritized test suite

# Empirical Results: Efficiency



Prioritizers are **useful** in practice because they incur **low time** overheads

# Concluding Remarks



Algorithms and Software Tools

Detailed Empirical Evaluations

- **Experimental** computer science involves the **implementation** and **evaluation** of algorithms that handle **real-world** problems
- Solving the **software crisis** with freely available **data sets** and free/open source **computational artifacts**

<http://www.cs.allegheny.edu/~gkapfham/research/>