

Hydra:

Automatically Configuring Algorithms for Portfolio-Based Selection

Lin Xu, Holger H. Hoos, Kevin Leyton-Brown



Department of Computer Science
University of British Columbia

Two automated algorithm design ideas

SATzilla

[Xu, Hutter, Hoos, Leyton-Brown, 2007; 2008]
portfolio-based algorithm selection



SATenstein

[KhudaBukhsh, Xu, Hoos, Leyton-Brown, 2009]
algorithm design via automatic configuration



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SATzilla

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portfolio-based algorithm selection



Exploit per-instance variation between solvers using learned runtime models

- **practical:** e.g., won 10 medals in 2007, 2009 SAT competitions
- **fully automated:** requires only cluster time rather than human design effort

Key drawback:

- requires a set of **strong, relatively uncorrelated** candidate solvers
- **can't be applied** in domains for which such solvers do not exist

Some particularly related work: [Rice, 1976]; [Leyton-Brown, Nudelman & Shoham, 2003; 2009]; [Guerri & Milano, 2004]; [Nudelman, Leyton-Brown, Shoham & Hoos, 2004]

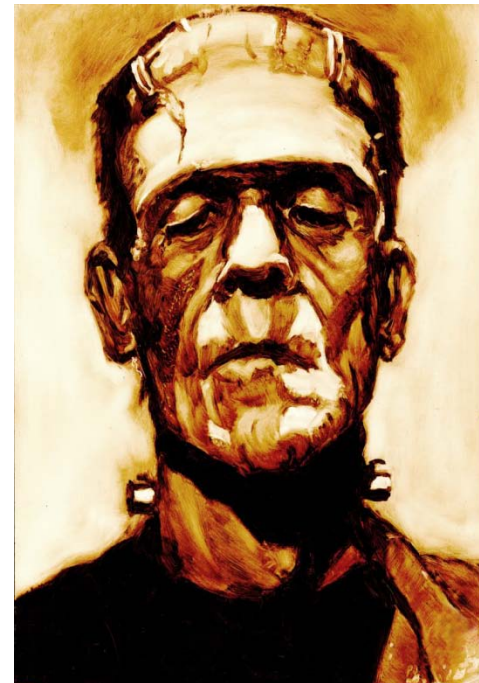
Two automated algorithm design ideas

SATenstein

[KhudaBukhsh, Xu, Hoos, Leyton-Brown, 2009]

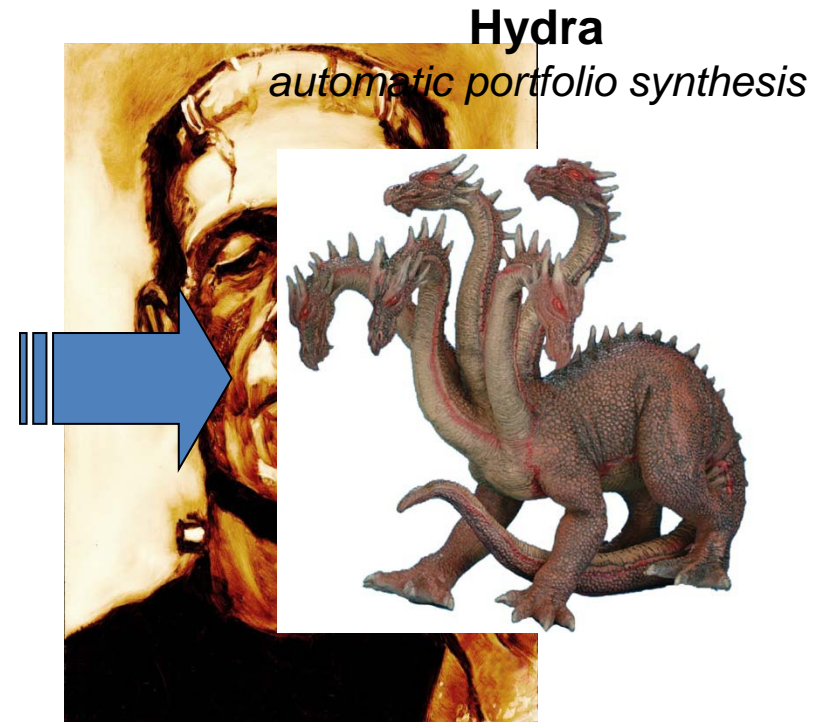
algorithm design via automatic configuration

- Instead of manually exploring a design space, build a **highly-parameterized algorithm** and then configure it automatically
- Can find **powerful, novel designs**
 - matched or outperformed existing SLS algorithms on six SAT domains
- But: **only produces single algorithms** designed to perform well on the entire training set



Some particularly related work: [Gratch & Dejong, 1992]; [Fukunaga, 2002]; [Balaprakash, Birattari & Stutzle, 2007]; [Hutter, Babic, Hoos & Hu, 2007]; [Ansotegui, Sellmann & Tierney, 2009]; [Hutter, Hoos, Stutzle & Leyton-Brown, 2009]

Two automated algorithm design ideas



Starting from a **single parameterized algorithm**, automatically find a set of **uncorrelated configurations** that can be used to build a **strong portfolio**.

Plan of This Talk

Background

- *SATzilla: Portfolio-Based Algorithm Selection*
- *SATenstein: Algorithm Configuration as Design*

Portfolio Synthesis

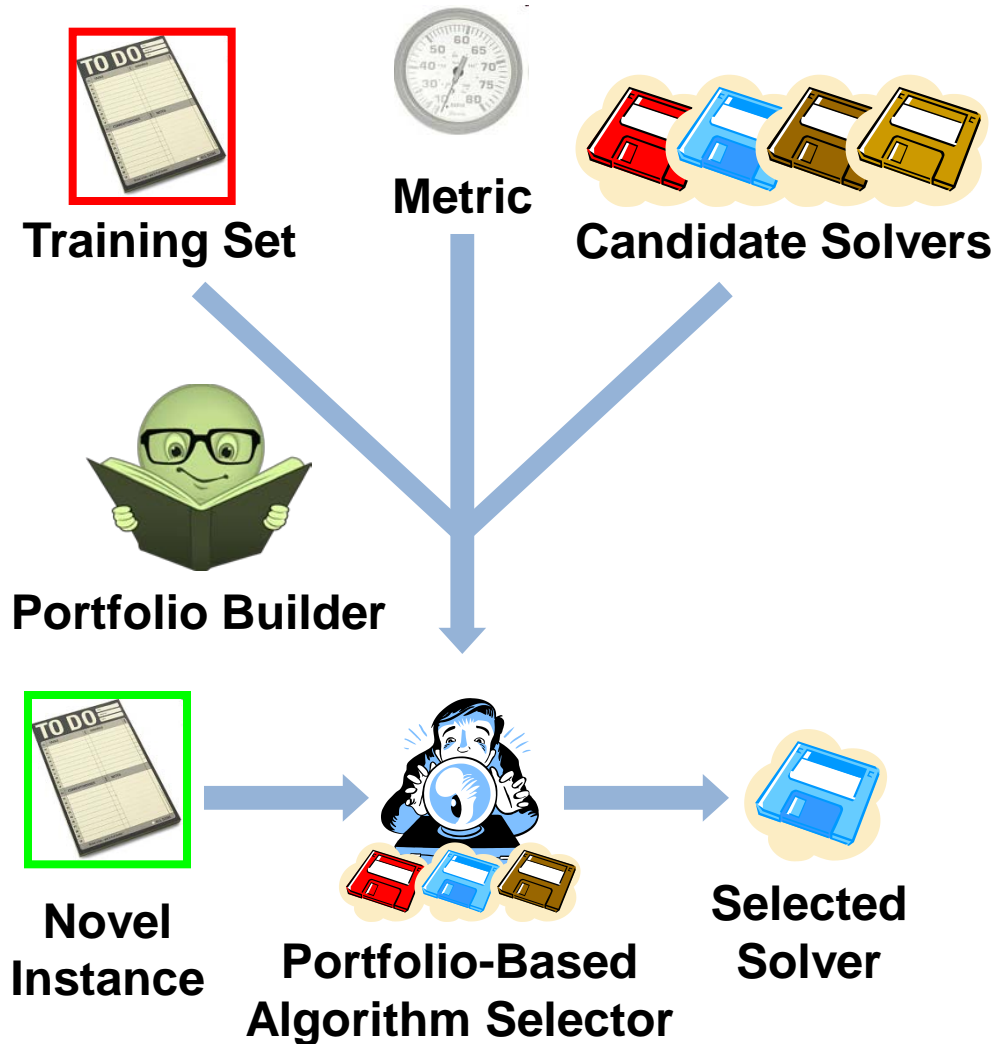
- *Related Work*
- *Hydra*

Experimental Results

Conclusions and Future Work

SATzilla: Portfolio-Based Algorithm Selection

[Xu, Hutter, Hoos, Leyton-Brown, 2007; 2008]



- **Given:**
 - training set of instances
 - performance metric
 - candidate solvers
 - portfolio builder (*incl. instance features*)
- **Training:**
 - collect performance data
 - portfolio builder learns predictive models
- **At Runtime:**
 - predict performance
 - select solver

SATenstein: Automated Algorithm Design

[KhudaBukhsh, Xu, Hoos, Leyton-Brown, 2009]



**Existing
Algorithm Components**



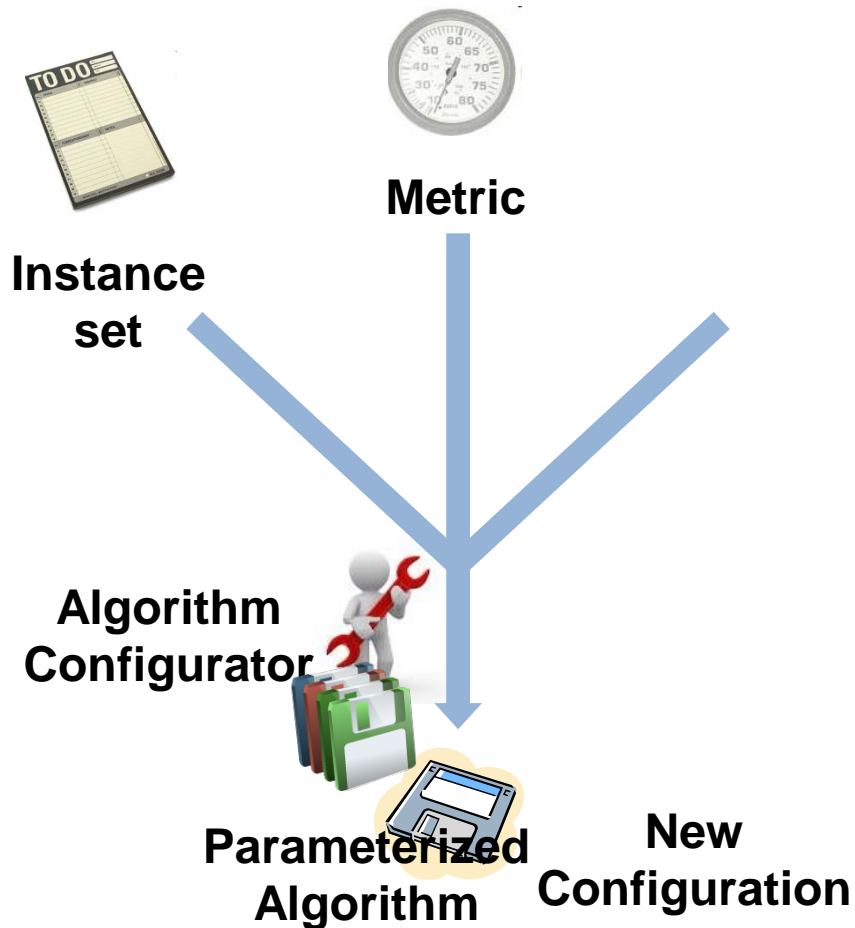
**Domain
Expert**



**Parameterized
Algorithm**

- **Designer creates highly-parameterized algorithm from existing components**
- **Given:**
 - training set of instances
 - performance metric
 - parameterized algorithm
 - algorithm configurator
- **Configure algorithm:**
 - run configurator on training instances
 - output is a configuration that optimizes metric

SATenstein: Automated Algorithm Design



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- *Related Work*
- *Hydra*

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Related Work

- Algorithm synthesis; portfolios and online algorithm selection
[Minton 1993]; [Huberman, Lukose & Hogg 1997]; [Howe et al, 1999]; [Gomes & Selman 2001]; [Carchrae & Beck 2005]; [Gagliolo & Schmidhuber 2006]; [Streeter, Golovin & Smith 2007]; [Roberts & Howe, 2007]; [Gaspero & Schaerf 2007]; [Monette, Deville & van Hentenryck 2009]
- **Two proposals for synthesis of selection-based portfolios:**
 1. **“Boosting as a Metaphor for Algorithm Design”** [L-B et al., 2003; 2009]
 2. **Stochastic Offline Programming** [Malitsky & Sellmann, 2009]
 - partition instances into k clusters based on features
 - find best-performing algorithm for each cluster
 - ⇒ assumes that all algorithms repeatedly (1) sample from a distribution over heuristics; (2) use the sampled heuristic for one search step
 - ⇒ best-performing algorithms identified using a custom optimization method
 - ⇒ *our goal is to construct an entirely general method for portfolio synthesis*
- **CP-Hydra** [O’Mahony, Hebrard, Holland, Nugent, & O’Sullivan, 2008]
 - selection-based portfolio for constraint programming

Boosting as a Metaphor for Algorithm Design

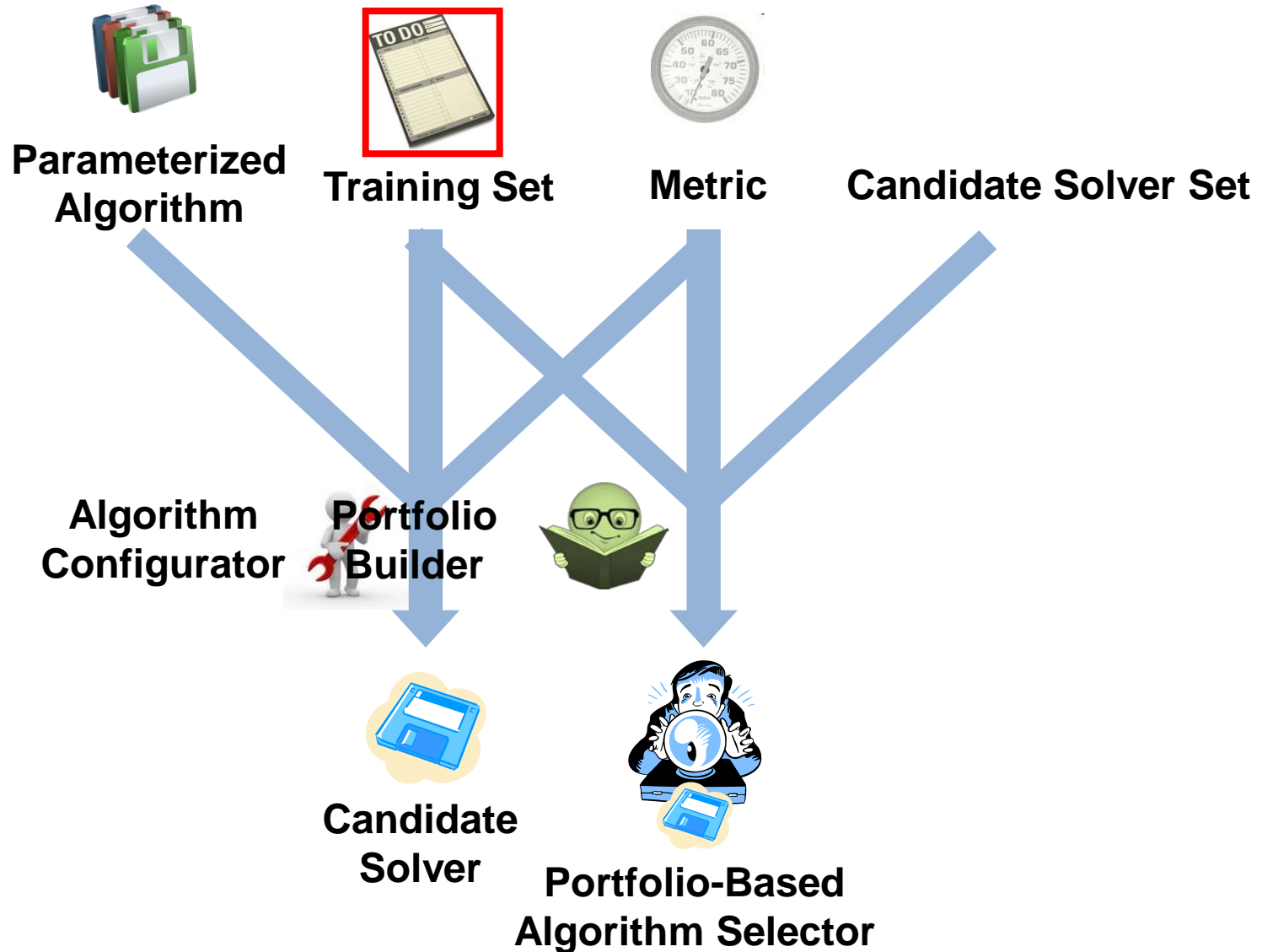
[Leyton-Brown, Nudelman, Andrew, McFadden, Shoham, 2003]; [Leyton-Brown, Nudelman, Shoham, 2009]

- **Core idea**
 - **re-weight instance distribution** to emphasize problems on which an existing portfolio P performs poorly
- **Interpretation as an automatic procedure:**
 - generate a new distribution D that is hard for P
 - find a new solver **maximizing average performance** on D
- **We intended to implement this procedure. But:**
 - discovered examples in which the algorithm with best average performance **does not improve the portfolio**
 - thus, the portfolio synthesis procedure can stagnate, even when other, helpful algorithms exist

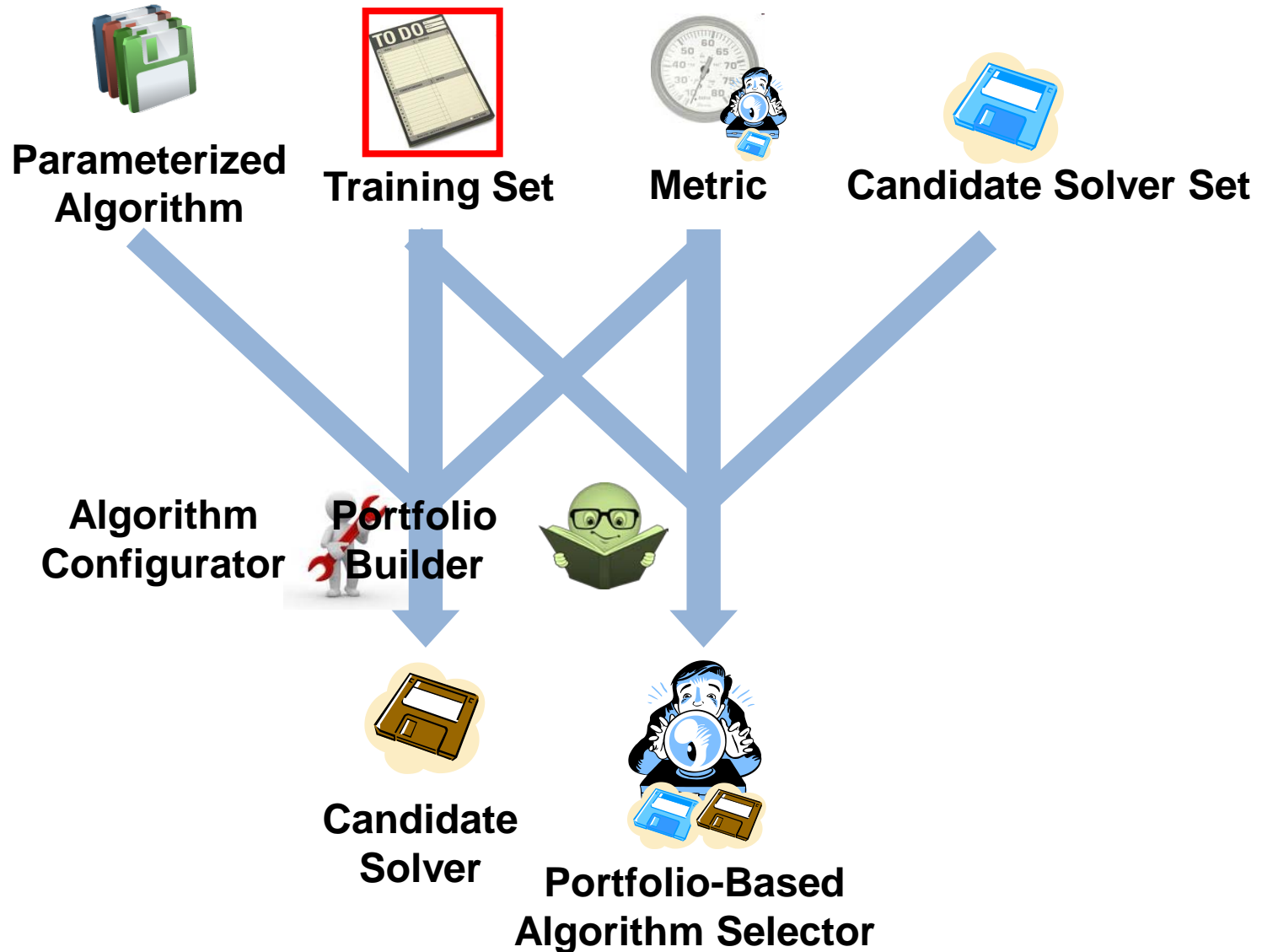
Hydra: Dynamic Performance Metric

- **Avoid stagnation via a dynamic performance metric:**
 - return performance of s when s outperforms P
 - return performance of P otherwise
- **Intuitively:** s is scored for its **marginal contribution** to P
- **This metric is given to an off-the-shelf configurator**, which optimizes it to find a new configuration s^*
- **Thus, we retain the same core idea as “boosting”**
 - build a new algorithm that explicitly aims to **improve upon an existing portfolio**
- **Contrast with Stochastic Offline Programming:**
 - algorithms target **sets of instances having very different features**
 - these feature differences can be irrelevant to algorithm performance

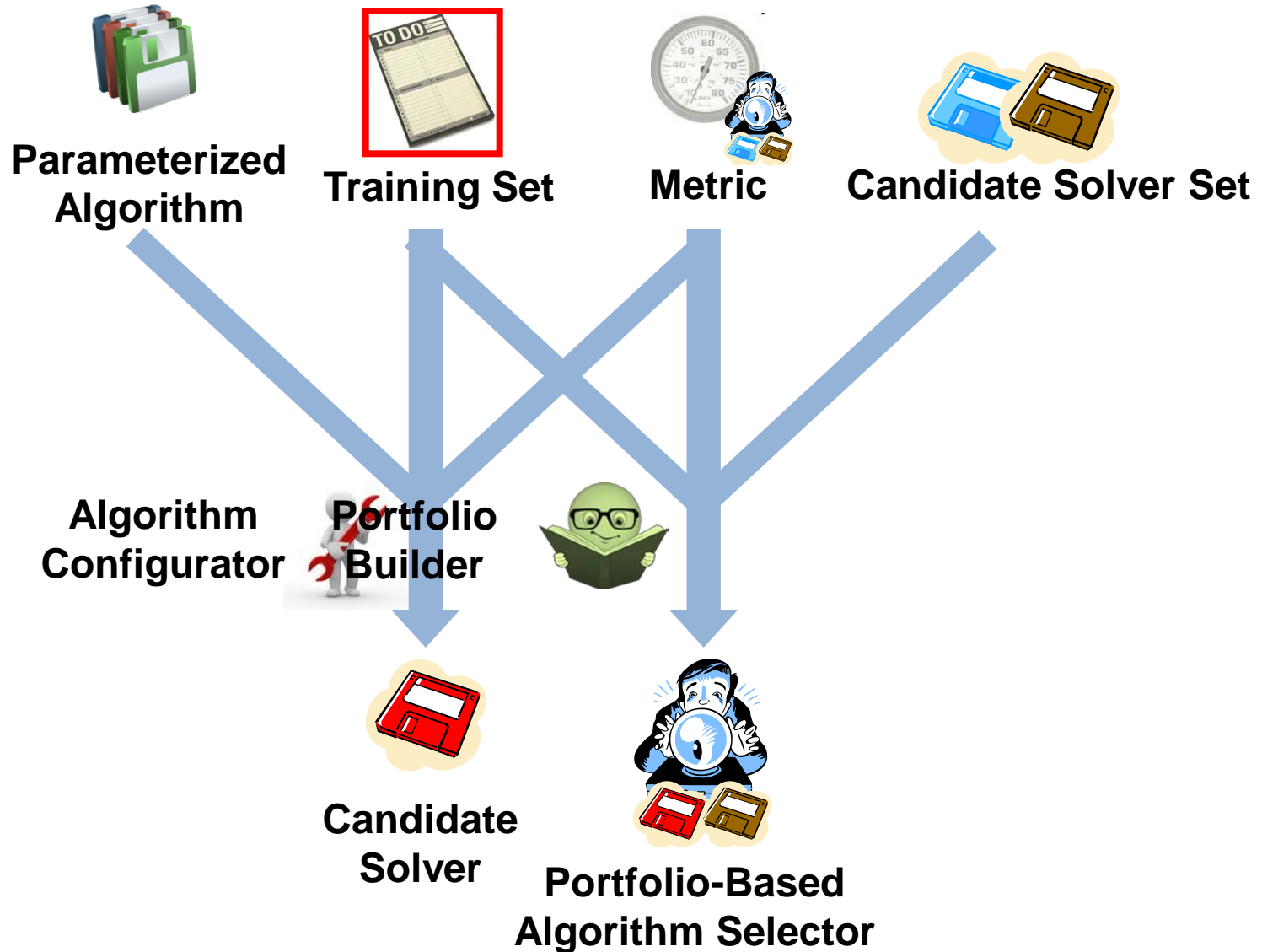
Hydra Procedure: Iteration 1



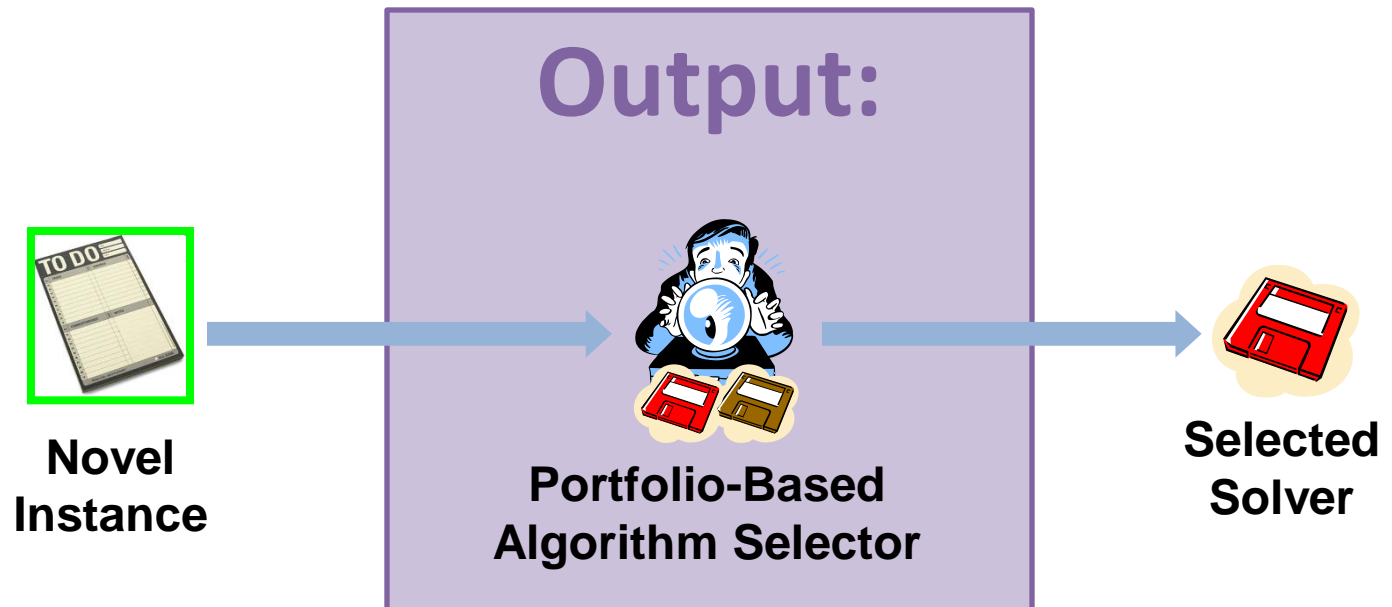
Hydra Procedure: Iteration 2



Hydra Procedure: Iteration 3



Hydra Procedure: After Termination



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Portfolio Synthesis

- *Related Work*
- *Hydra*

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Problem Domain

- **Even though Hydra is most useful in other domains, we evaluated it on SAT.**
- **High bar for comparison**
 - strong state-of-the-art solvers
 - portfolio-based solvers already successful
 - ⇒ to be able to argue that Hydra does well,
we want to compare to a strong portfolio
- **Pragmatic benefits**
 - a wide variety of interesting datasets
 - existing instance features
 - SATenstein is a suitable configuration target

Experimental Setup: Hydra's Inputs



- **Portfolio Builder:** SATzilla framework

[Xu, Hutter, Hoos, Leyton-Brown, 2008]



- **Parameterized Solver:** SATenstein-LS

[KhudaBukhsh, Xu, Hoos, Leyton-Brown, 2009]



- **Algorithm Configurator:** FocusedILS 2.3

[Hutter, Hoos, Leyton-Brown, 2009]



- **Performance Metric:** Penalized average runtime (PAR)

- **Instance Sets:**

- ❑ 2 from SATenstein paper

[KhudaBukhsh, Xu, Hoos, Leyton-Brown, 2009]

- ❑ 2 from previous SAT competitions



Experimental Setup: Challengers

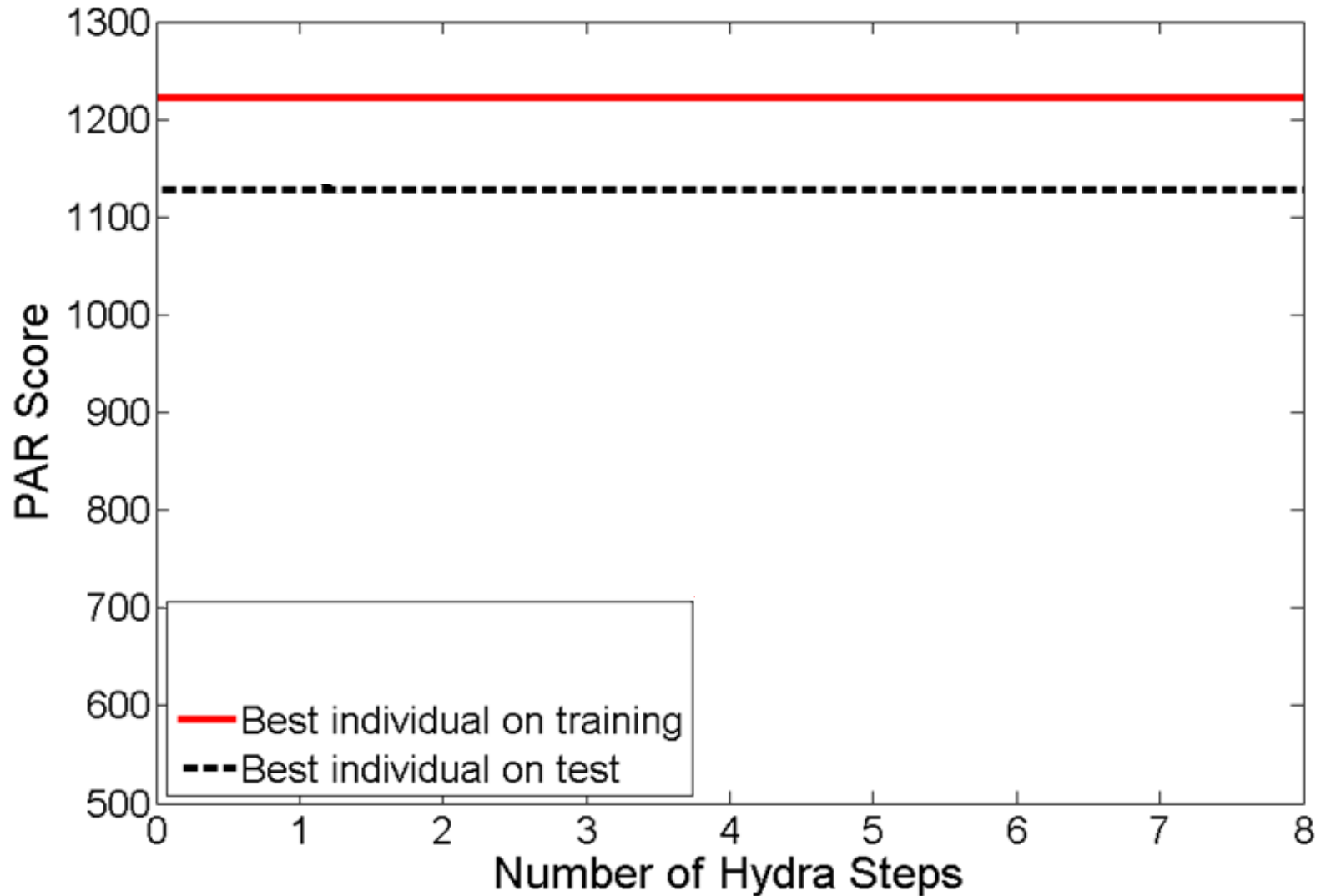
- **Individual state-of-the-art solvers**
 - 11 **manually-crafted** SLS solvers
 - all 7 SLS winners of any SAT competition 2002 – 2007
 - 4 other prominent solvers
 - 6 **SATenstein solvers**
- **Also considered portfolios of challengers**
 - used same **portfolio builder** (SATenstein)

Performance Summary

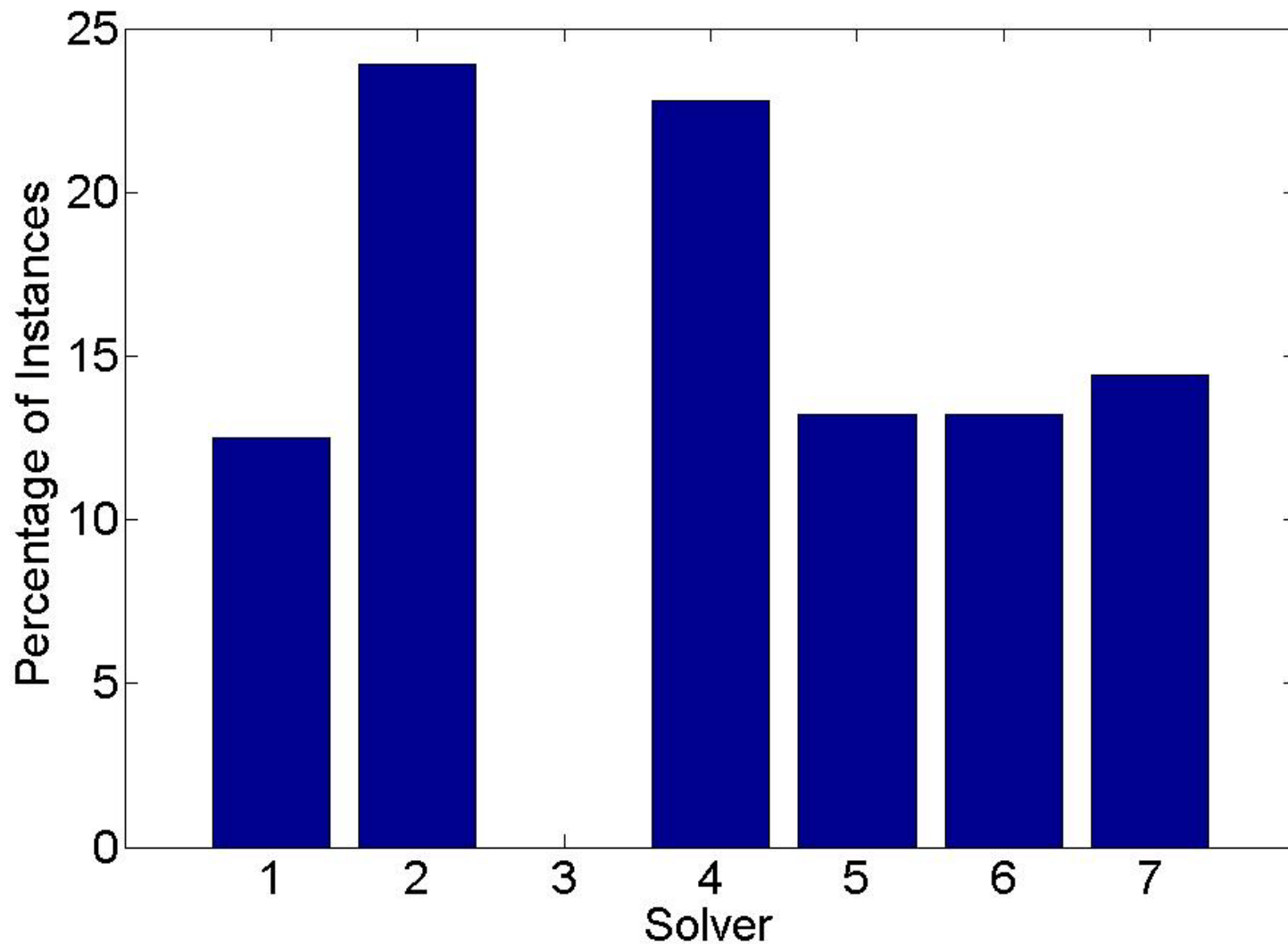
Solver	RAND	HAND	BM	INDU
Best Challenger (of 17)	1128.63	2960.39	224.53	11.89

** Statistically insignificant performance difference (sign rank test).
Hydra's performance was significantly better in all other pairings.*

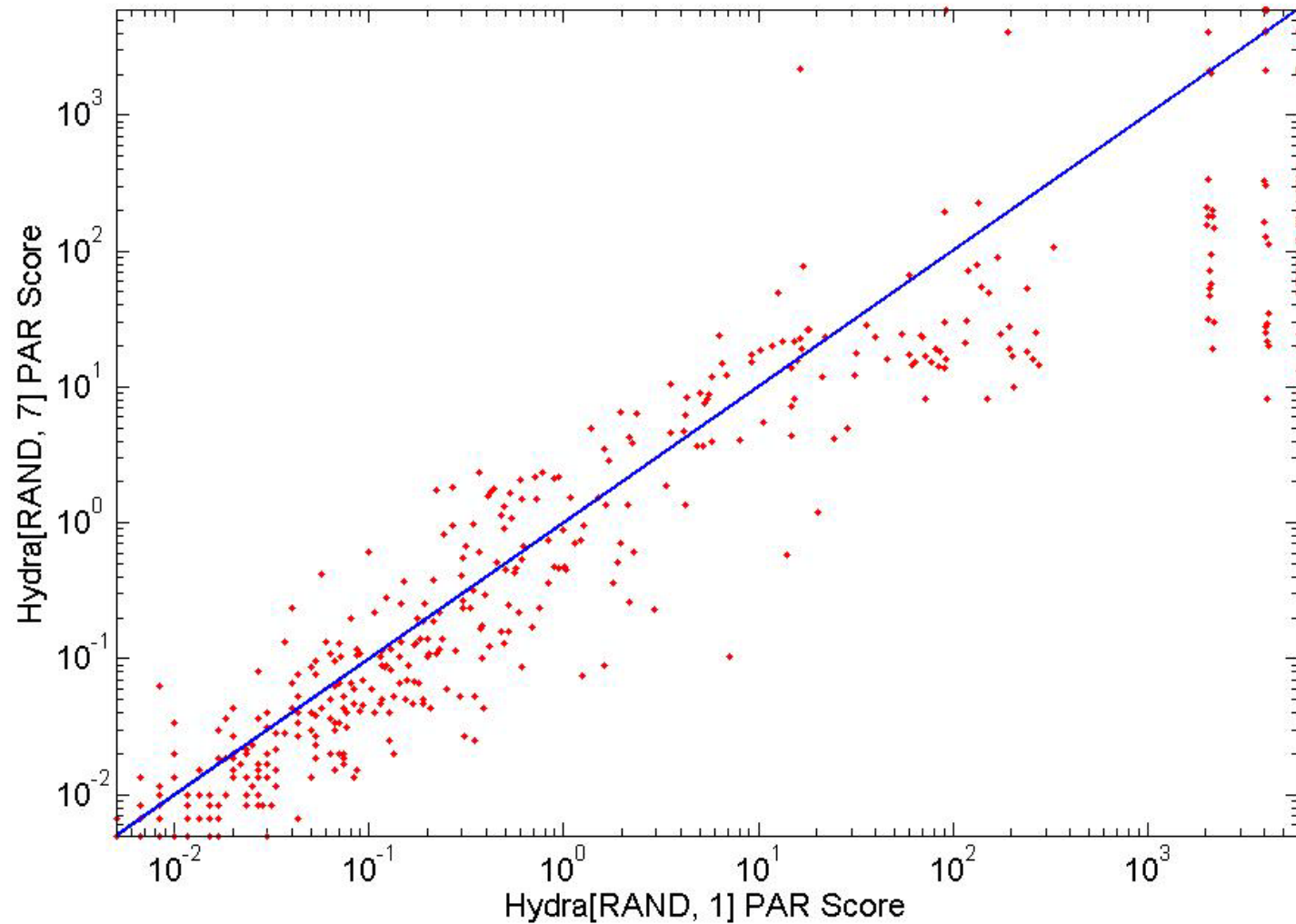
Performance Progress, RAND



Selection Percentages After 7 Iterations, RAND



Improvement After 7 Iterations, RAND



Conclusions

- **Hydra: an automatic design approach combining**
 - portfolio-based algorithm selection (here: “SATzilla”)
 - automated algorithm configuration (here: “SATenstein”)
- Completely automated
- Algorithm/configurator/portfolio-builder agnostic
- Most useful in domains where few strong solvers exist
- Nevertheless met or exceeded state-of-the-art performance on SLS for SAT in 4 domains

Thank You!