Predicting job running time in DIRAC

Master project

DIRAC user community meeting

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Presentation Outline

- Introduction to my Master project
- Predator
  - Architecture and design
  - Challenges
  - Other systems
  - Technology
  - Project outlook
- Summary
- Conclusion

Master project

Goal:

Predict the job running time given the job configuration and site binding.

Project Status:

Not implemented

Motivation for the project

- Production users
  - Want to know when their jobs are finished
- Improve scheduling
  - More accurate resource reservations
  - Resource selection

Limitations

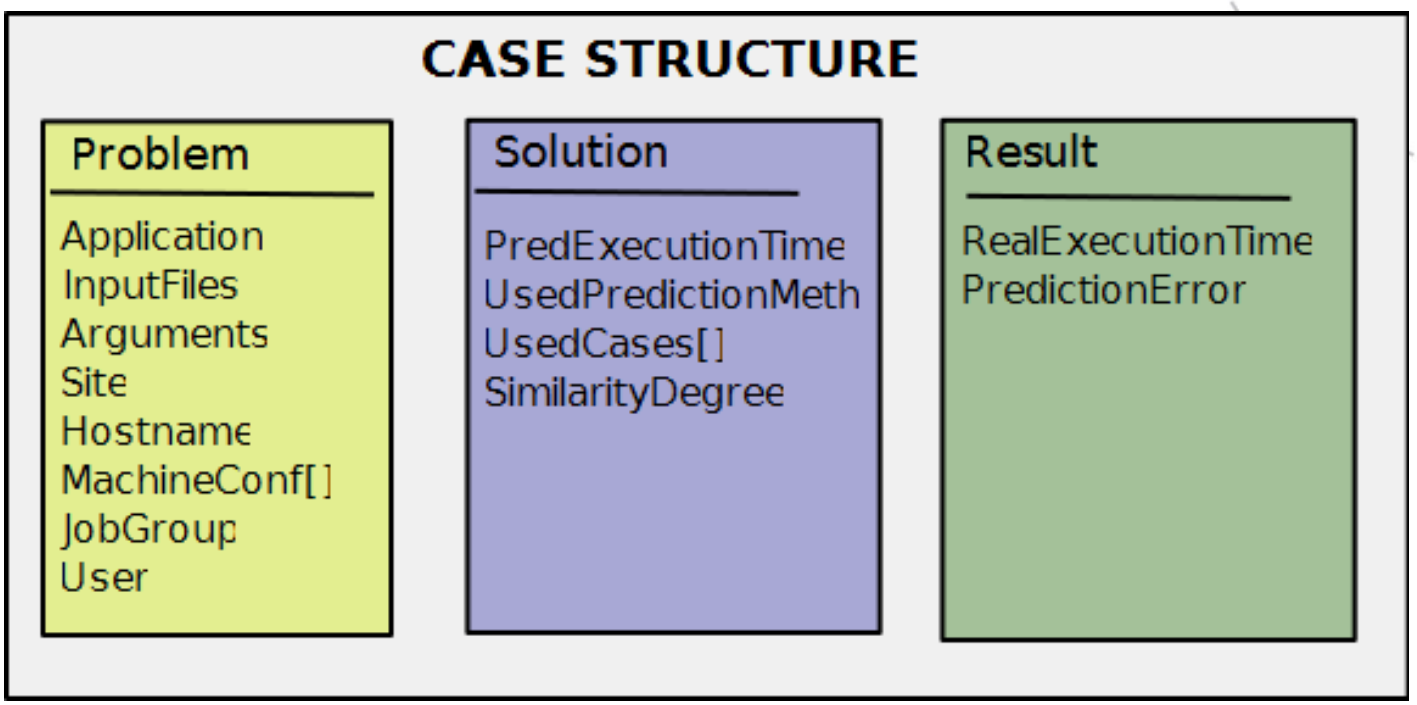
Will only consider production jobs:

- Generation
- Simulation
- Reconstruction

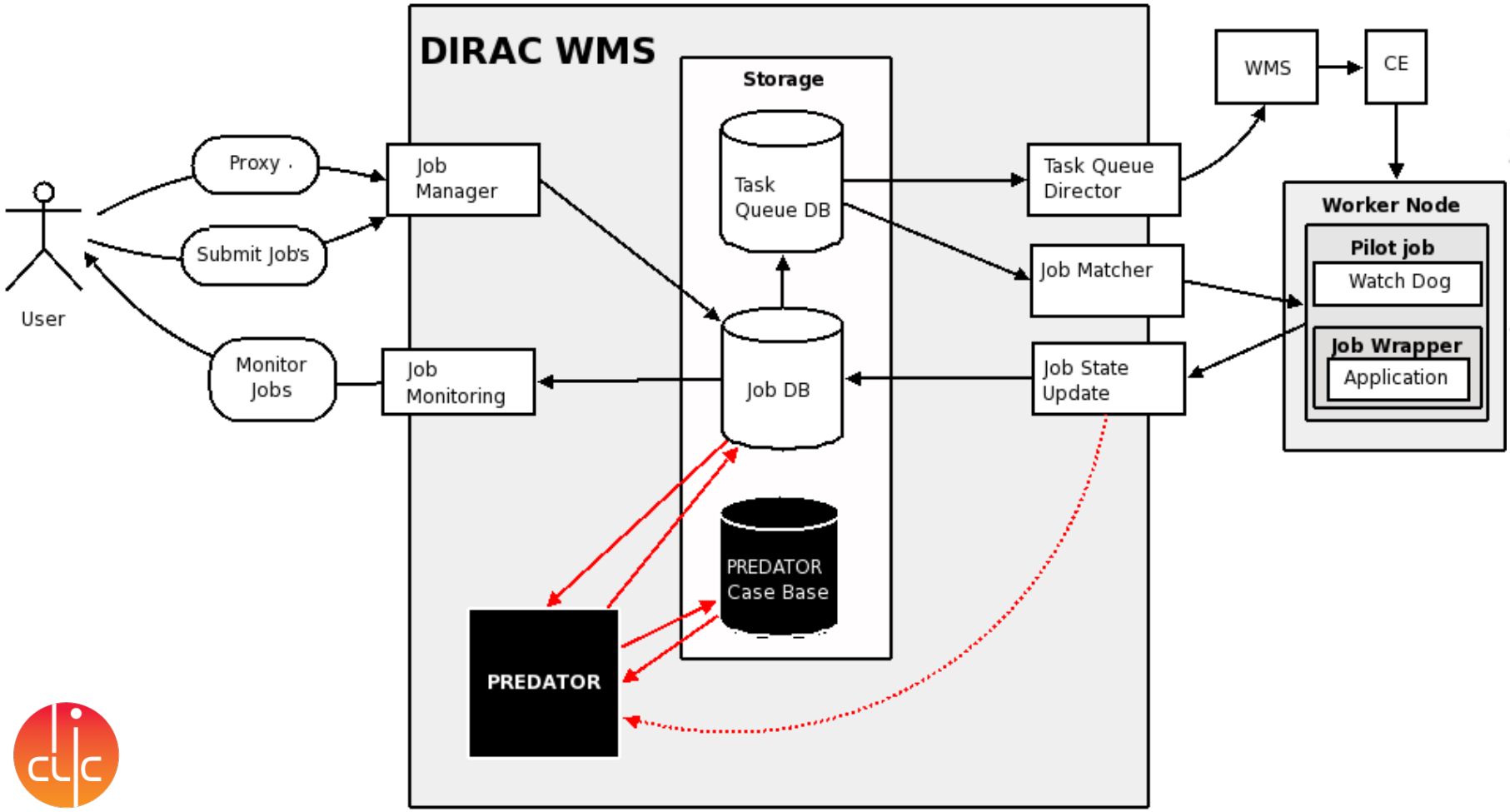
PREDATOR

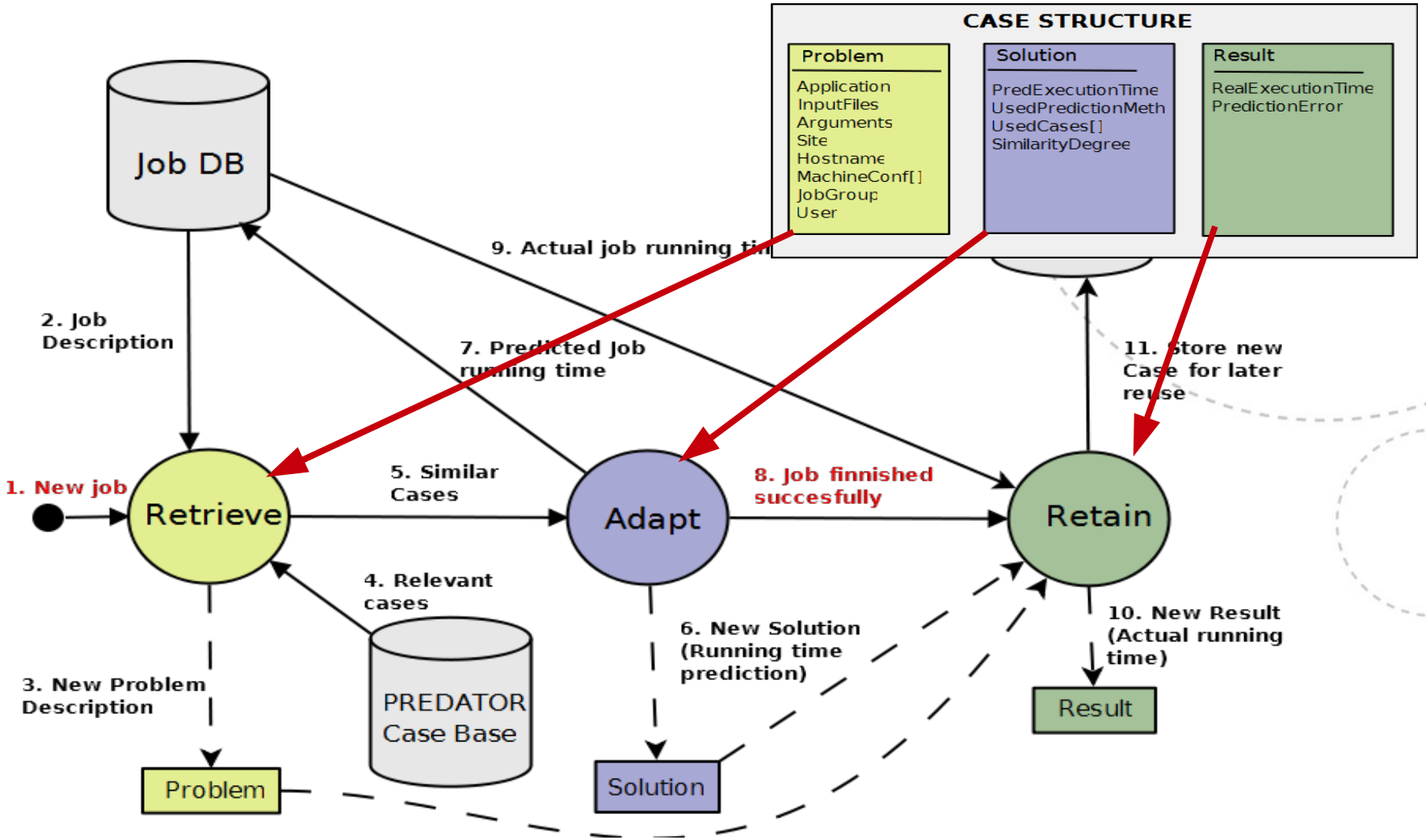
PREDATOR

- Use Case Based Reasoning
  - Inspired by how humans solve problems
  - Learning is the process of solving problems and storing them for later reuse
  - Solve new problems by combining previous experiences
  - The knowledge of the system is the case base


CASE STRUCTURE
Problem
Application
InputFiles
Arguments
Site
Hostname
MachineConf[]
JobGroup
User
Solution
PredExecutionTime
UsedPredictionMeth
UsedCases[]
SimilarityDegree
Result
RealExecutionTime
PredictionError

PREDATOR and DIRAC WMS

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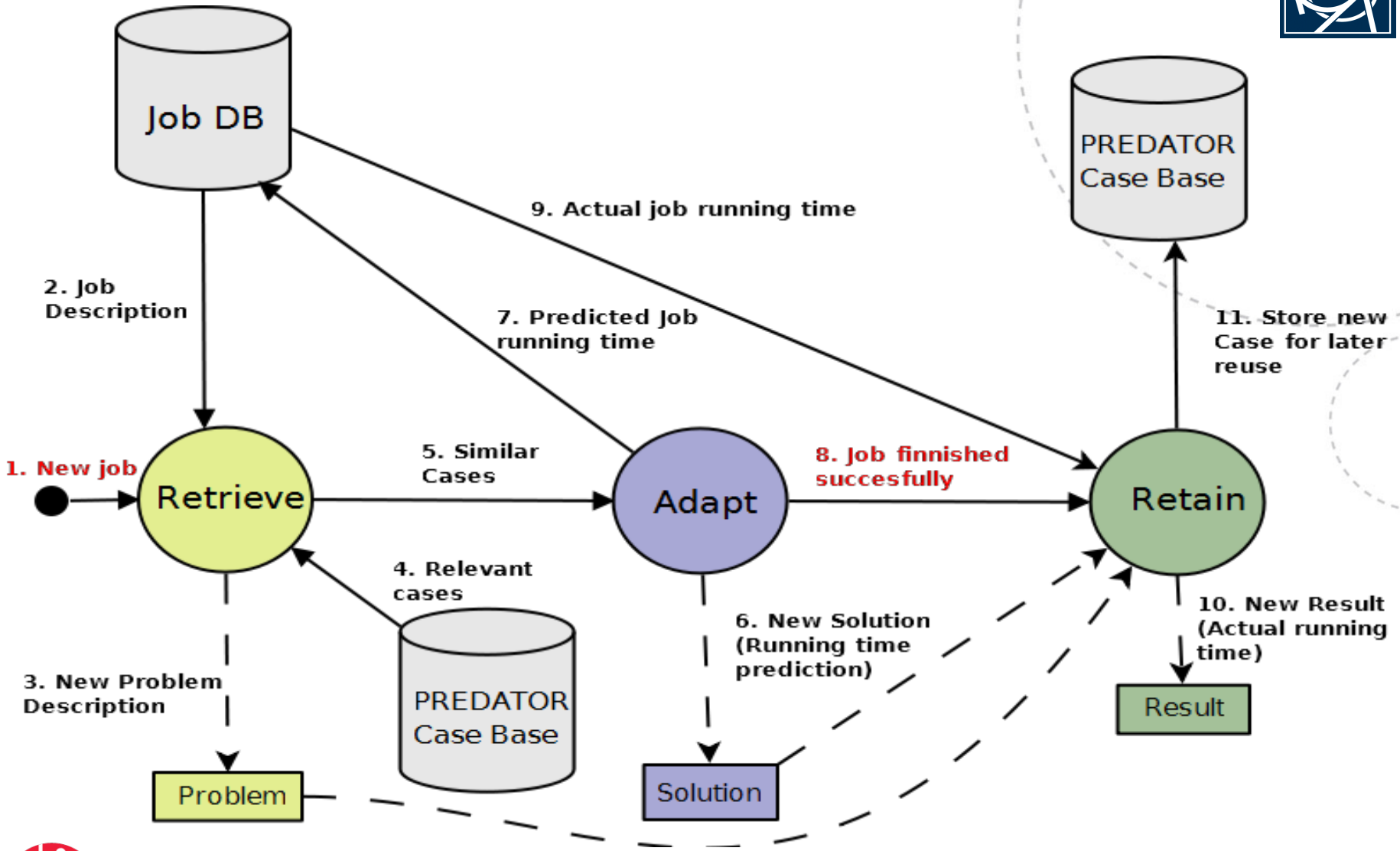
Job DB
9. Actual job running ti
2. Job Description
7. Predicted Job running time
11. Store new Case for later reuse
1. New job
Retrieve
5. Similar Cases
Adapt
8. Job finnished succesfully
Retain
4. Relevant cases
PREDATOR Case Base
6. New Solution (Running time prediction)
10. New Result (Actual running time)
3. New Problem Description
Result
Problem
Solution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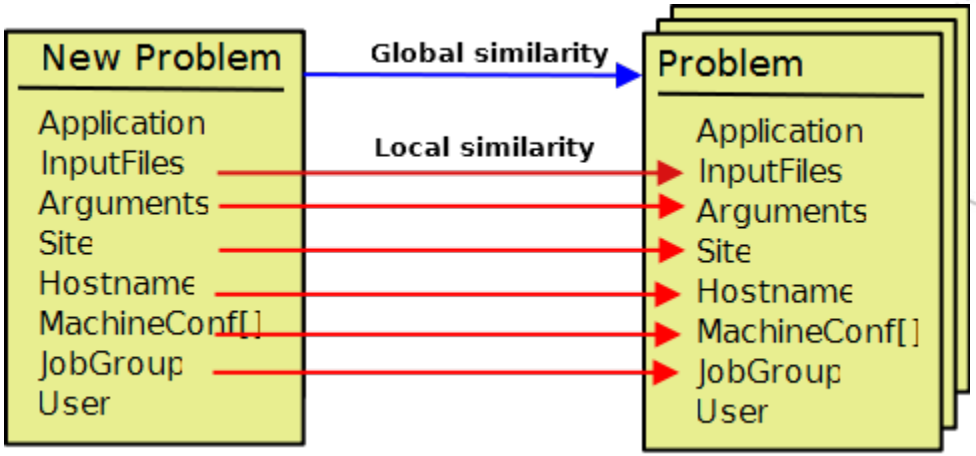
Retrieve

1. Retrieve job description from JobDB
2. Create new problem instance
3. Build relevance vector (E=[Site,Application,...])
4. Retrieve relevant cases from case base into memory
5. Apply similarity measurements
6. Select the most similar cases

Retrieve cont. Similarity measurements

Local similarity functions

- Numeric: $\text{sim}(a, b) = 1 - \frac{|a - b|}{range}$
- Taxonomy: $\text{sim}(a, b) = \frac{h(\text{common node}(a, b))}{\min(h(a), h(b))}$
- Symbolic: $\text{sim}(a, b) = \begin{cases} 1, & \text{if } a = b \\ 0, & \text{if } a \neq b \end{cases}$
- Multi-valued: $\text{sim}(a, b) = \frac{\text{card}(a) \cap \text{card}(b)}{\text{card}(a \cup b)}$



Global similarity function

$$\text{sim}(A, B) = \sum_{i=1}^{p} \omega_i \text{sim}_i(a_i, b_i)$$

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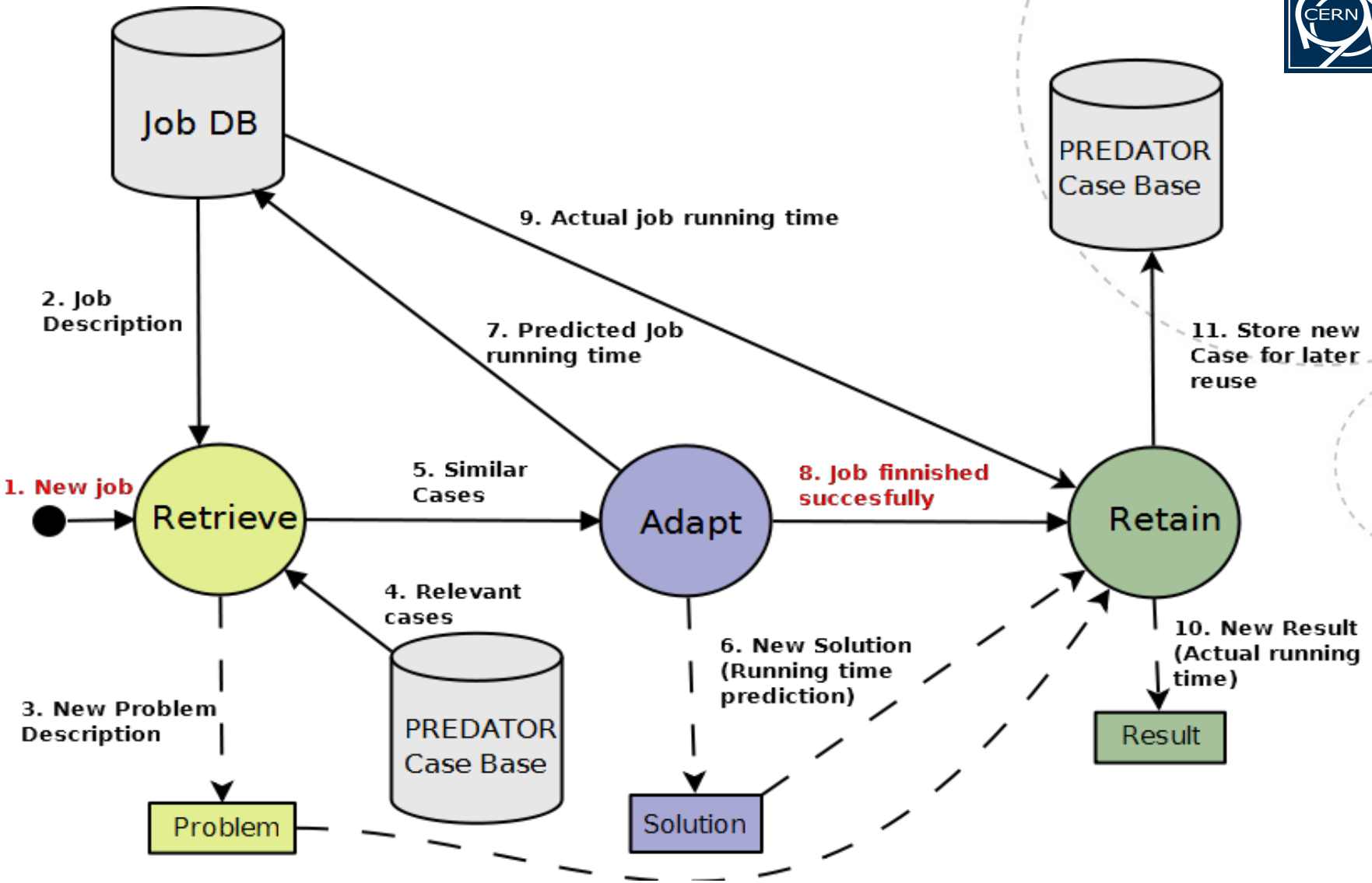
Adapt

- From the retrieved cases, average the real execution time of the result instances -> predicted running time
- Can also use different prediction methods based on which prediction method was most succesful in the past

Retain

- Upon job completion, running time is fetched from Job DB
- Result object is created and the prediction error is calculated
- The Problem, Solution, Result is stored in the Case Base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What challenges are we facing?

- The conditions on the worker node is not the same every time
- Heterogeneity within a GRID site
- Jobs are submitted in bunches → Bottlenecks

Other systems

- **PredCase**: CBR used to determine in which resource to run the job.
  - 9% Average prediction error
- **I. Foster et.al** sorts similar jobs into buckets, the predicted runtime is an average of the running time of the jobs in the bucket.

$$PredictionError = \frac{PredictedTime - MeasuredTime}{MeasuredTime} \times 100\%$$

Technology

- jCOLIBRI CBR framework (>= Java 1.6)
  - Need to define case structure
  - Simularity mecanism
  - Reuse\combine how the cases are combined to create the new solution
- Protege with myCBR to implement simularity funcion, can be exported and imported into jCOLIBRI

Project outlook

- Results available end of June
- Thesis by the end of July

Suggestions

- Accounting data → GRID workloads archieve
  - Standard: the Grid Workload Format
- GridSim → A Grid Simulation Toolkit

"The goal of the Grid Workloads Archive is to provide a virtual meeting place where practitioners and researchers can exchange grid workload traces. "

Summary

- PREDATOR: Predicting running time of production jobs in DIRAC
  - Uses CBR
- Technology: jCOLIBRI, Protege, myCBR
- GRID workload archieve, standard for grid workload traces

Conclusion

- PREDATOR must be implemented and tested
- Results should be available by end of june
- There are other systems that have shown this concept is possible

References

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  In Proceedings of the Workshop on Job Scheduling Strategies for Parallel Processing
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  In Proceedings of the 2004 IEEE International Symposium on Cluster Computing and the Grid