



# Parameter Estimation for Human Trust in Information Sources

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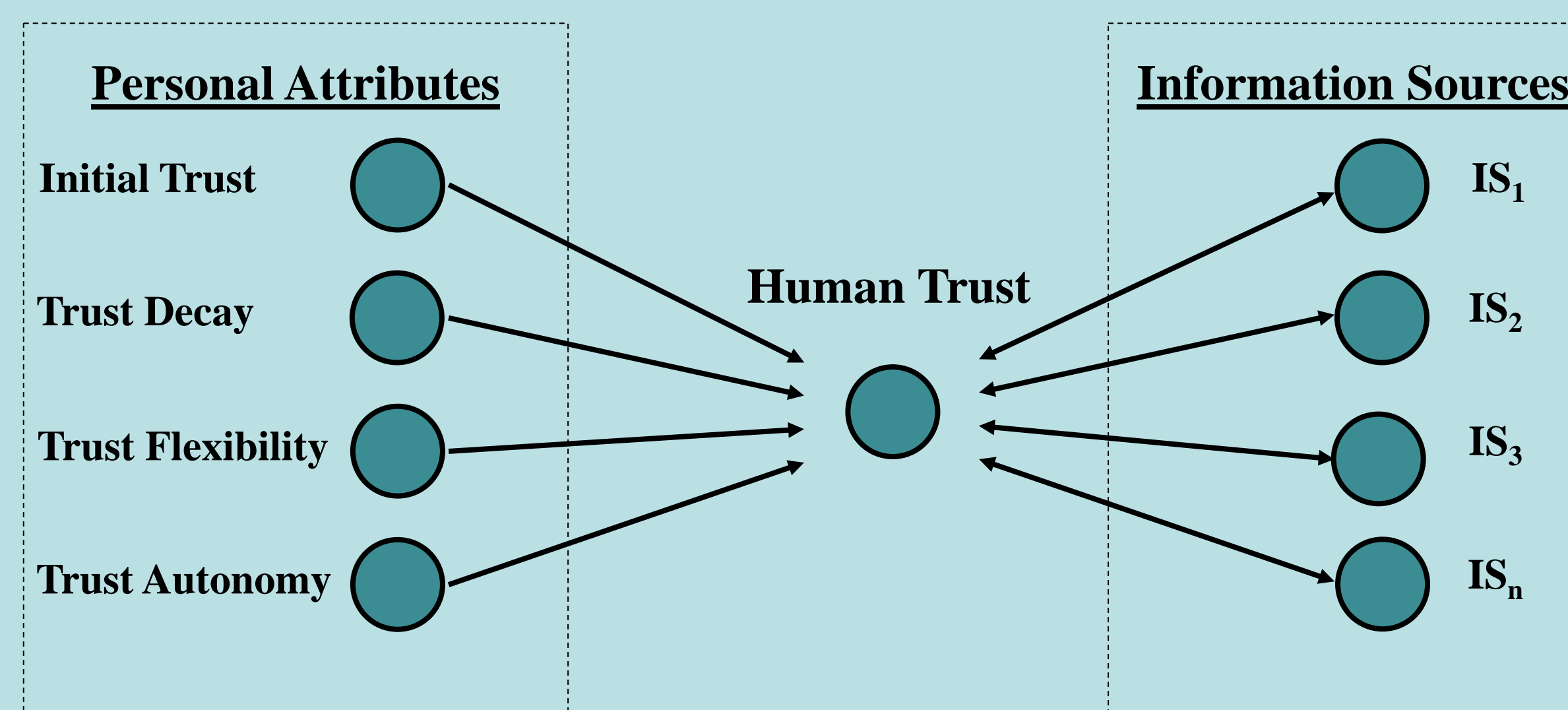
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## PROBLEM

- The behavior of a computational human trust model crucially depend on the specific values of different parameters related to human personality.
- How these parameters can be tuned to a specific individual?

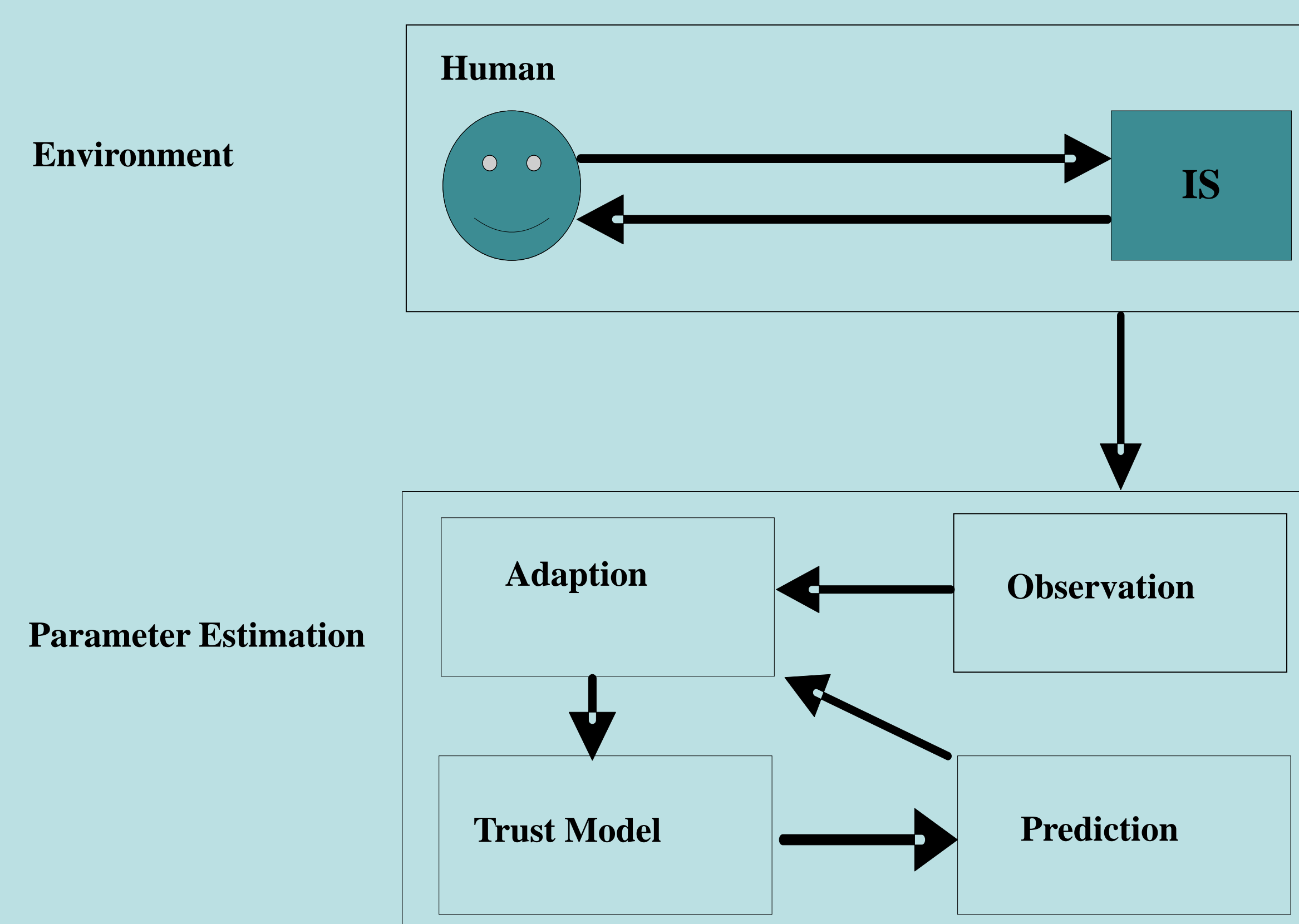
## EXAMPLE MODEL [1]

- Human Personality Attribute
  - Initial bias (initial trust)
  - Memory (trust decay)
  - Experience Frugality (trust flexibility)
  - Dependence of Information Sources (trust autonomy)
- At each time step human requests one of the competitive information source with high trust value.



Dynamic Relationships in the Human Trust Model on Competitive Information Sources

## APPROACH



Parameter Estimation for Human Trust in Information Sources

## ACCURACY

Accuracy = Correct Predictions / Observed Behaviors

## METHODS

### EXHAUSTIVE SEARCH METHOD

- On each observation entire attribute search space is explored to find the vector of parameter settings with maximum accuracy.
- Guarantees the optimal solution.

### BISECTION SEARCH METHOD

- On each observation entire attribute search space is explored by halving the intervals for the parameter values to find the vector of parameter settings with maximum accuracy.
- Does not guarantee the optimal solution.

### EXTENDED BISECTION SEARCH METHOD

- Find parameter setting with maximum accuracy as in Bisection Search Method
- Store parameters in list of solutions for future use.
- Compare the accuracy of current parameter setting with all known solutions in the list.
- Does not guarantee the optimal solution.

### STIMULATED ANNEALING METHOD

- Search the parameter with maximum accuracy in the neighborhood of initial parameters selected randomly.
- Neighborhood for search depends on computational budget and accuracy of parameter setting  
 $Neighborhood = ComputationalBudgetLeft * (1 - Accuracy)$

## TIME COMPLEXITY

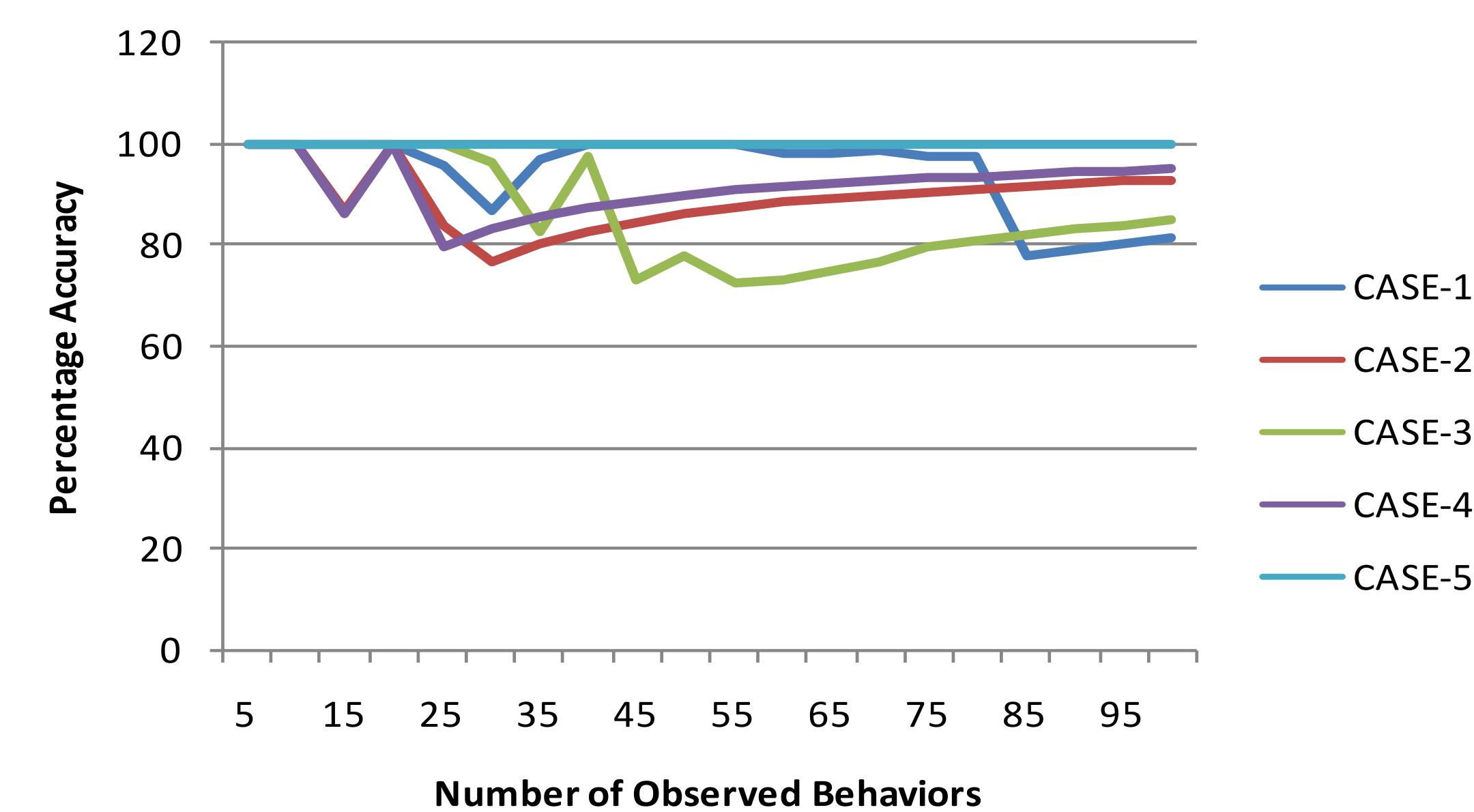
| Methods                   | Complexity                  |
|---------------------------|-----------------------------|
| Exhaustive Search         | $O((10)^{\alpha\tau} NB^2)$ |
| Bisection Search          | $O(\alpha\tau NB^2)$        |
| Extended Bisection Search | $O(\alpha\tau NB^2)$        |
| Simulated Annealing       | $O(CNB^2)$                  |

$\alpha$  = No. of Parameters,  $\tau$  = Precision of Estimation, N = No. of Information Sources, B = Observed Behaviors, C = Computational Budget (for stimulated annealing)

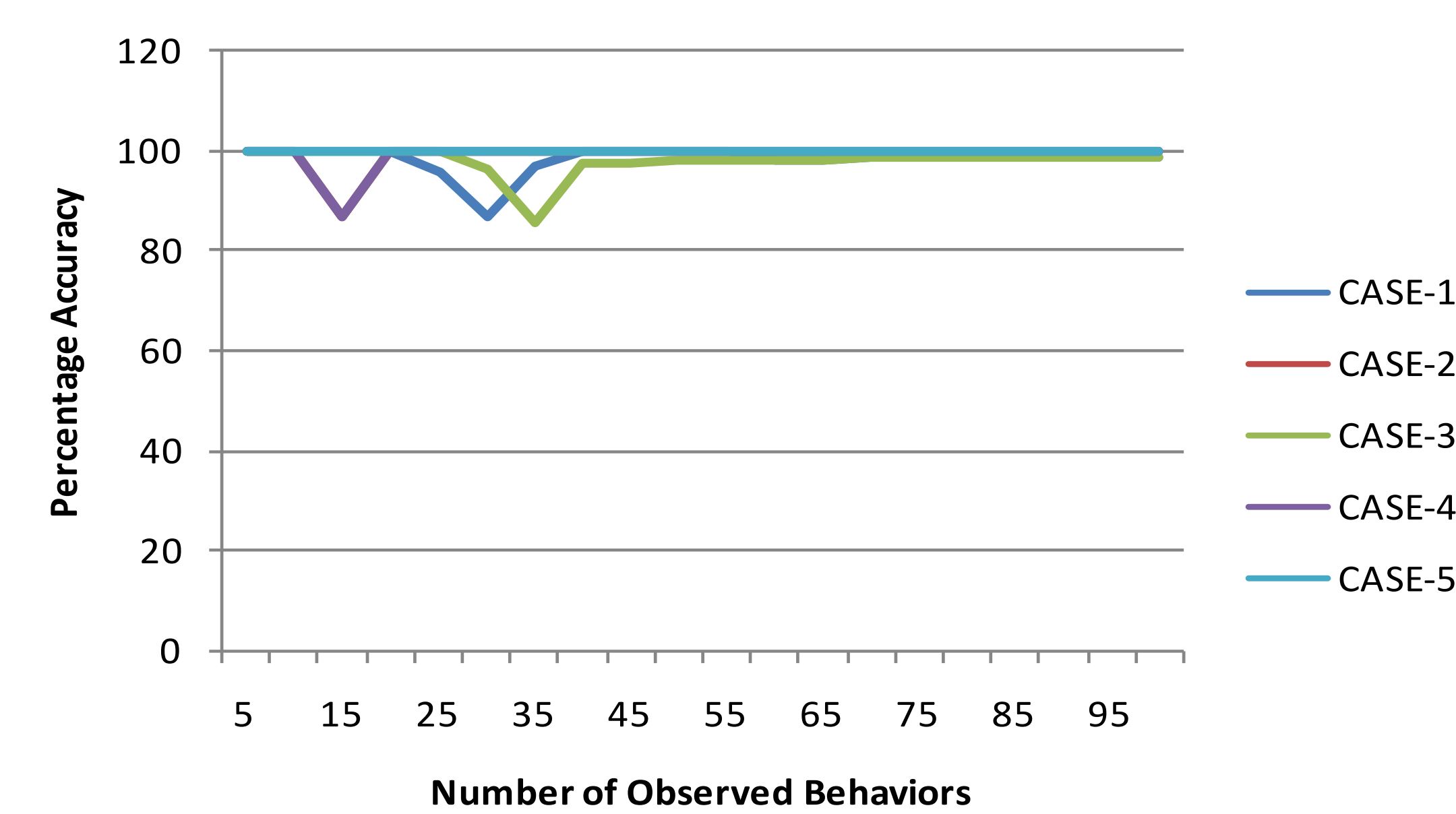
## SIMULATION AND RESULTS

### Experimental Configurations for generating Human Behavior

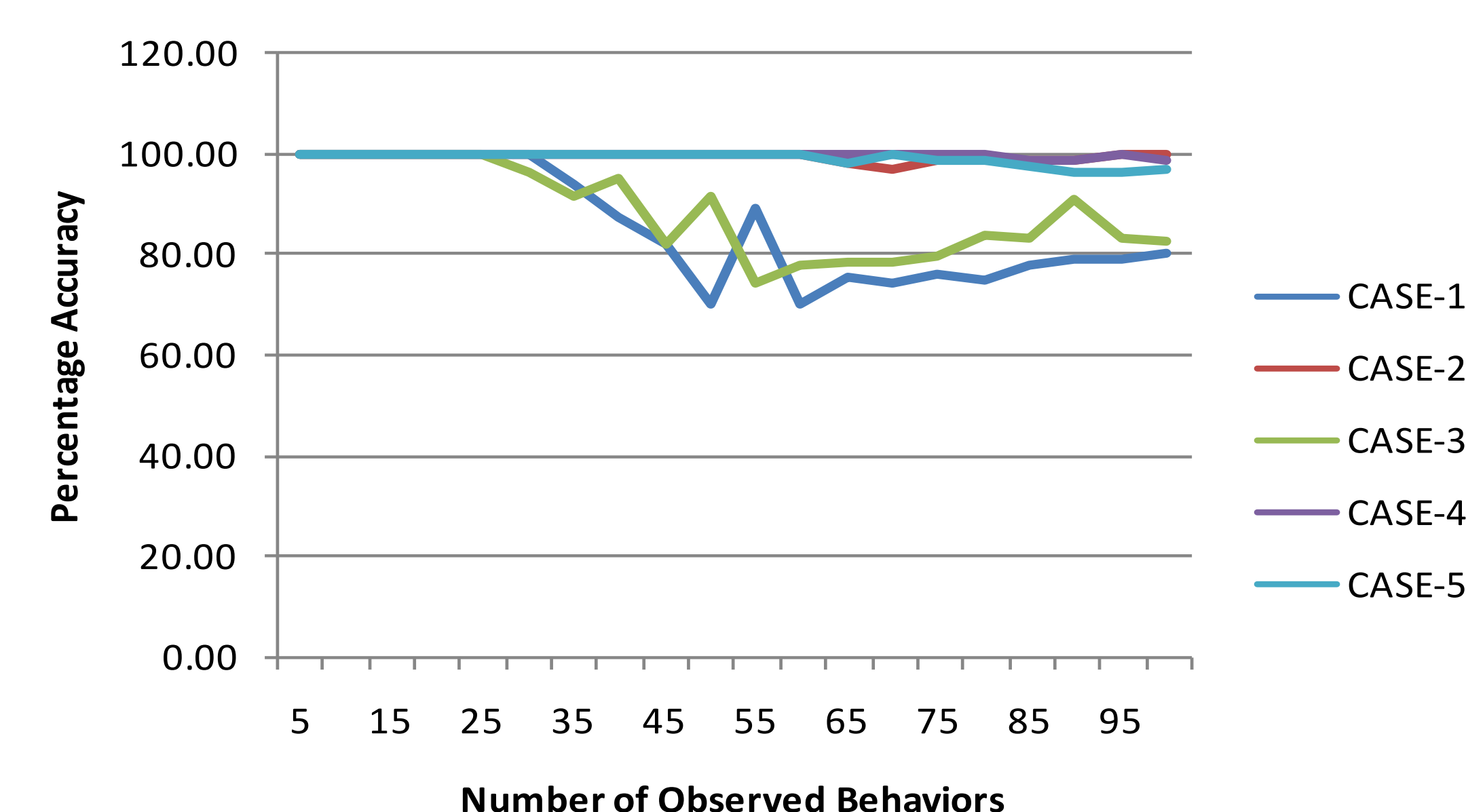
| Case   | 1                | 2                | 3                | 4                | 5                |
|--|------------------|------------------|------------------|------------------|------------------|
| No. of Parameters  | 3                | 3                | 3                | 3                | 3                |
| Precision (digits)   | 2                | 2                | 2                | 2                | 2                |
| Information Sources  | 3                | 3                | 3                | 3                | 3                |
| IS <sub>1</sub> , IS <sub>2</sub> , IS <sub>3</sub> Responses                      | 1,-1,-1          | 1,-1,-1          | 1,-1,-1          | 1,-1,-1          | 1,-1,-1          |
| Observed Behaviors   | 100              | 100              | 100              | 100              | 100              |
| Trust Decay  | 0.01             | 0.01             | 0.01             | 0.25             | 0.01             |
| Trust Flexibility  | 0.75             | 0.75             | 0.25             | 0.75             | 0.75             |
| Trust Autonomy   | 0.25             | 0.25             | 0.25             | 0.25             | 0.75             |
| Human Initial Trust on Sources IS <sub>1</sub> , IS <sub>2</sub> , IS <sub>3</sub> | 0.00, 0.15, 0.30 | 0.00, 0.05, 0.10 | 0.00, 0.05, 0.10 | 0.00, 0.05, 0.10 | 0.00, 0.05, 0.10 |



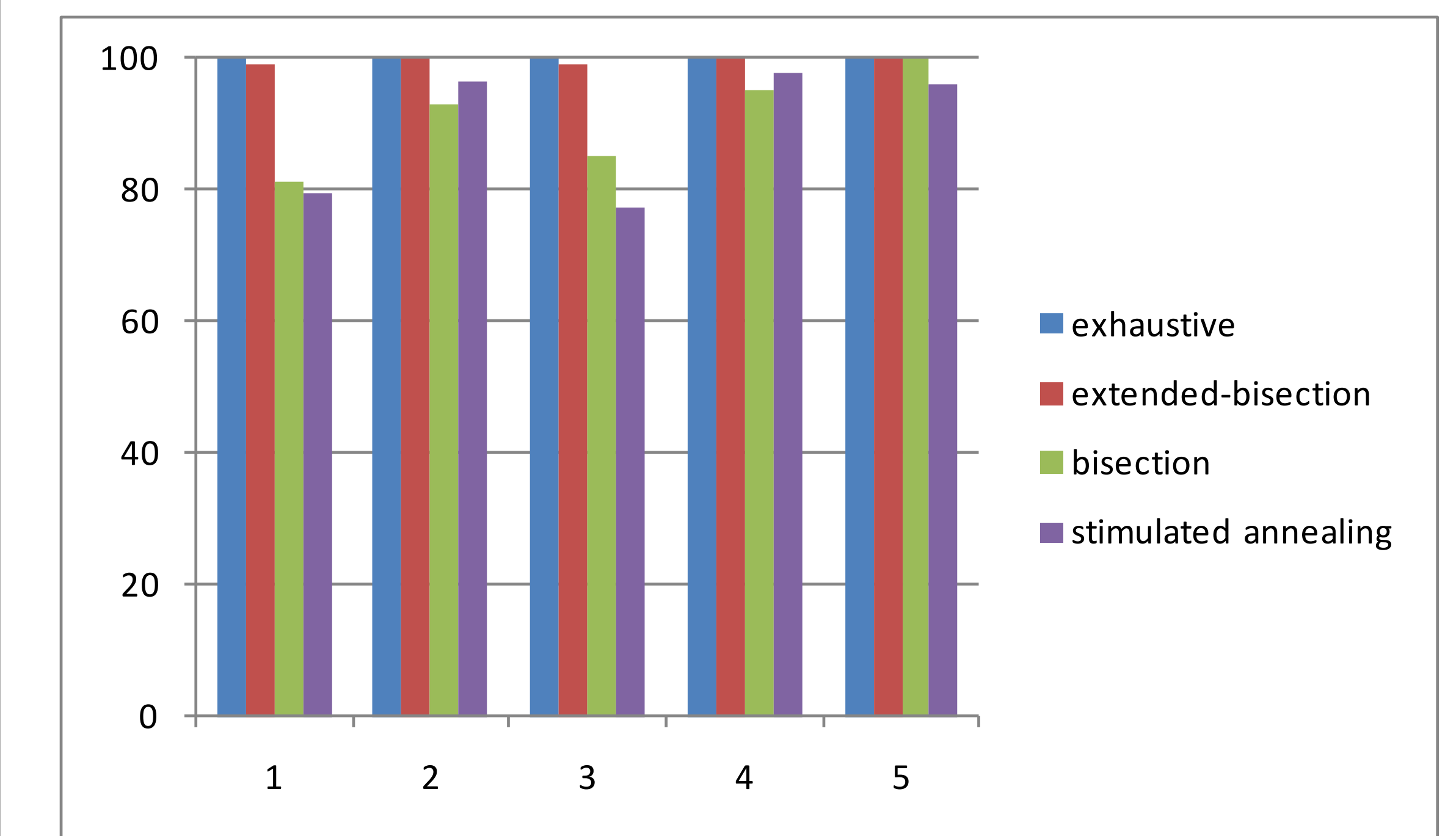
Accuracy of Bisection Search Method



Accuracy of Extended Bisection Search Method



Maximum Percentage Accuracy of Stimulated Annealing



Accuracy of All Methods in Case 1, 2, 3, 4, and 5

## FORMAL ANALYSIS

### P1(trace, λ): Information reduces possible solutions

If at time point t the set of observed behaviors is of size n and the number of solutions found is p1, then if at a later point in time t2 the set of observed behaviors is larger (i.e. m > n), then the number of solutions found p2 is less than or equal to p1 \* λ.

### Formally:

$\forall t:time, n:integer, p1:integer$

$[state(\gamma, t) \models number\_observed\_behaviors(n) \ \& \ state(\gamma, t) \models number\_solutions(p1)] \Rightarrow [\forall t2:time, m:integer, p2:integer \ [ [t2 > t \ \& \ m > n \ \& \ state(\gamma, t2) \models number\_observed\_behaviors(m) \ \& \ state(\gamma, t2) \models number\_solutions(p2)] \Rightarrow p2 \leq p1 * \lambda ] ]$

## CONCLUSIONS

- Computation time of the exhaustive search scales up worst, whereas the Simulated Annealing approach scales up best.
- Exhaustive search finds the most accurate settings, whereas Simulated Annealing sometimes only comes up with poor solutions.
- The bisection, and extended bisection are right in the middle: They do have a higher accuracy and are computationally less expensive.
- The choice of which method to use ultimately depends on the domain.

## REFERENCE

[1] Hoogendoorn, M., Jaffry, S.W., and Treur, J., "Modeling Dynamics of Relative Trust of Competitive Information Agents", In: Klusch, M., Pechoucek, M., Polleres, A. (eds.), Proceedings of the 12th International Workshop on Cooperative Information Agents, CIA'08. LNAI, vol. 5180. Springer, 2008, pp. 55-70.