Towards Comparing and Combining Points-to Analyses

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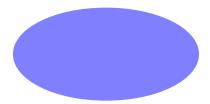
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- Static dataflow analysis
- Computes reference information, e.g., possible targets of a call and possible objects referenced by a field (in the following, we use abstract *result sets* for illustration)
- Input to, e.g., optimizing compilers, software analysis tools
- Requirements: accuracy, speed

- Different implementations use different data structures
- Different papers use different metrics
 - Hard to tell, how they compare to each other
- If we can properly *compare* analysis results from different implementations, we can also *combine* them, thus exploiting the "best of" from different approaches/implementations

Gold standard (G): The "exact" result set of a given analysis

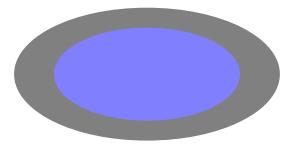


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Types of Analysis

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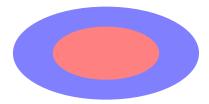
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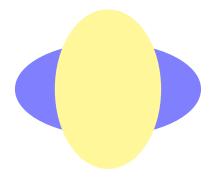


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Gold standard (G): The "exact" result set of a given analysis

- conservative analysis: over-approximation of G
- optimistic analysis: under-approximation of G
- general analysis: a mix of conservative and optimistic analysis



- Static analysis should be conservative
- This is often the case only for *subsets* of a programming language!
 - Dynamic class loading
 - Reflection
- Maybe even on purpose, to improve performance
- Applications:
 - Software understanding tools
 - Optimistic optimizations (with guard)

- Precision P: how much of the analysis result set is in G?
- Recall R: how much of G is found through the analysis?
- Accuracy F (F-score): Weighted measure between Precision and Recall

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Special case of	Comparison of A_1, A_2 with respect to		
A_1, A_2	P	R	F-score
A_1, A_2 cons.	$P_1 \ge P_2 \Leftrightarrow A_1 \le A_2 $	$R_1 = R_2 = 1$	$F_1 \ge F_2 \Leftrightarrow A_1 \le A_2 $
A_1, A_2 opt.	$P_1 = P_2 = 1$	$ R_1 \ge R_2 \Leftrightarrow A_1 \ge A_2 $	$F_1 \ge F_2 \Leftrightarrow A_1 \ge A_2 $
A_1 cons. A_2 opt.	$P_1 \leq P_2 = 1$	$R_1=1\geq R_2$	$F_1 \ge F_2 \Leftrightarrow \frac{ A_1 }{ G } \le \frac{ G }{ A_2 }$
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- Two conservative analyses: Compute intersection
- Two optimistic analyses: Compute union

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What happens when improving a static analysis?

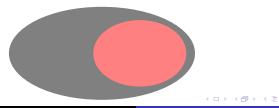
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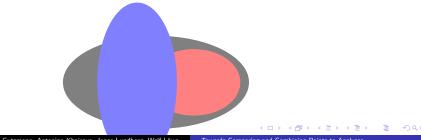
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- improve its precision (e.g., by context sensitivity) result set becomes smaller



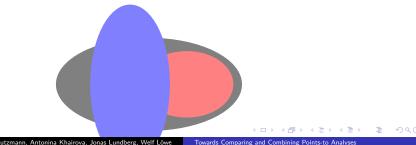
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- Assume we knew the Gold Standard
- $\bullet \rightarrow$ We cannot assess the benefits of an improved analysis for a non-conservative baseline analysis



• Points-to analyses for Java:

- Spark from the Soot-framework; inclusion-based, context-insensitive
- *Points-to SSA*; context-insensitive (CI) and two context-sensitive variants (CS₁ and CS₂) (SCAM 2008)
- Compare the results with results from dynamic analysis
- A set of 12 benchmark programs:
 - for 4 of them, all points-to analyses are conservative
 - for one of them, Spark is not conservative (missing support for a native method)
 - for the rest, all points-to analyses are general

• Combining conservative analyses:

- Spark and CI
- $\bullet~\mbox{CS}_1$ and \mbox{CS}_2
- measurable, yet very small improvements

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 - Yield (unspecified) approximations of precision, recall, accuracy
 - Improving analyses: When going from CI to CS₁ and CS₂
 - $\bullet\,$ In one case, a method that is reachable in the dynamic analysis and identified as such in CI, is no longer identified as reachable in CS_1 and CS_2
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 - For some fine-grained metrics, we also find "more misses"
- Optimistic vs. general analysis: For one project, the result sets of many metrics are bigger for the optimistic analysis than for the static ones

• Combining Analyses not worth it in practice

- Static analysis is not always conservative (although often assumed)
- Comparing two analyses wrt. accuracy is possible only in special cases, or when a Gold Standard is at hand
- When improving a non-conservative baseline analysis, we need to be very careful to interpret the results
- Sometimes, dynamic analysis has bigger result sets than static analysis
 - \rightarrow strictly more accurate

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