Integrating Induction and Deduction in Model Checking

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Model Checking and Synthesis

E. M. Clarke and E. A. Emerson, 1981:

"We propose a method of constructing concurrent programs in which the *synchronization skeleton of the program is automatically synthesized* from a high-level (branching time) Temporal Logic specification."

(1st sentence of their seminal model checking paper)

Three Messages in this Talk

[Seshia DAC'12; Jha & Seshia, SYNT'14]

- 1. Verification by Reduction to Synthesis
 - Many (verification) tasks involve synthesis
- 2. Induction + Deduction + Structure: An Effective Approach to Synthesis:
 - Induction: Learning from examples
 - Deduction: Logical inference and constraint solving
 - Structure: Hypothesis on syntactic form of artifact to be synthesized
 - "Syntax-Guided Synthesis" [Alur et al., FMCAD'13]
 - Inspired by Counterexample-guided abstraction refinement (CEGAR) [Clarke et al., CAV'00]

3. Machine Learning Theory + Formal Methods

- Analysis of Counterexample-Guided Synthesis
 - Sample Complexity, Convergence, Search Strategies __3_

Artifacts Synthesized in Verification

Inductive invariants

- Abstraction functions / abstract models
- Auxiliary specifications (e.g., pre/post-conditions, function summaries)
- Environment assumptions / Interface specifications
- Interpolants
- Ranking functions
- Intermediate lemmas for compositional proofs
- Theory lemma instances in SMT solving
- Patterns for Quantifier Instantiation



Formal Verification as Synthesis

Inductive Invariants

Abstraction Functions

One Reduction from Verification to Synthesis

NOTATION Transition system M = (I, δ) Safety property Ψ = G(ψ)

VERIFICATION PROBLEM Does M satisfy Ψ ?

SYNTHESIS PROBLEM Synthesize ϕ s.t. $I \Rightarrow \phi \land \psi$ $\phi \land \psi \land \delta \Rightarrow \phi' \land \psi'$

Two Reductions from Verification to Synthesis

NOTATION Transition system M = (I, δ), S = set of states Safety property Ψ = G(ψ)

VERIFICATION PROBLEMDoes M satisfy Ψ ?

SYNTHESIS PROBLEM #1 Synthesize ϕ s.t. $I \Rightarrow \phi \land \psi$ $\phi \land \psi \land \delta \Rightarrow \phi' \land \psi'$ SYNTHESIS PROBLEM #2 Synthesize $\alpha : S \rightarrow \hat{S}$ where $\alpha(M) = (\hat{I}, \hat{\delta})$ s.t. $\alpha(M)$ satisfies Ψ iff M satisfies Ψ

Counterexample-Guided Abstraction Refinement is "Inductive" Synthesis



Lazy SMT Solving performs "Inductive" Synthesis (of Lemmas)



CEGAR & Lazy SMT perform Active Learning from (Counter)Examples

Difference from standard learning theory: Learning Algorithm and Verification Oracle are typically "General" Solvers, independent of concept class



Machine Learning Theory ←→ Formal Methods: 2 Sample Results

- Lower Bounds on Convergence of Counterexample-guided loop
 - Teaching Dimension (TD): Minimum number of (labeled) examples a teacher must reveal to uniquely identify any concept from a class
 - <u>Thm:</u> TD is a lower bound on # iterations for counterexample-guided synthesis
- Impact of "Quality" of Counterexamples
 - Does the type of counterexample affect convergence for *infinite-size* concept classes?
 - <u>Thm:</u> Minimum counterexamples are no better than arbitrary counterexamples

Conclusion

- Model Checking and Synthesis: many connections
- Verification by Reduction to Synthesis
- Approach for Synthesis: Induction + Deduction + Structure
 - "Syntax-Guided Synthesis" [Alur et al., FMCAD'13]
 - Inspired by Counterexample-guided abstraction refinement (CEGAR) [Clarke et al., CAV'00]
- Machine Learning Theory & Formal Methods: theoretical connections
 - Sample Complexity, Convergence, Search Strategies