

Model-Based Specification of Complex Bidding Strategies in Agent-Based Online Auctions

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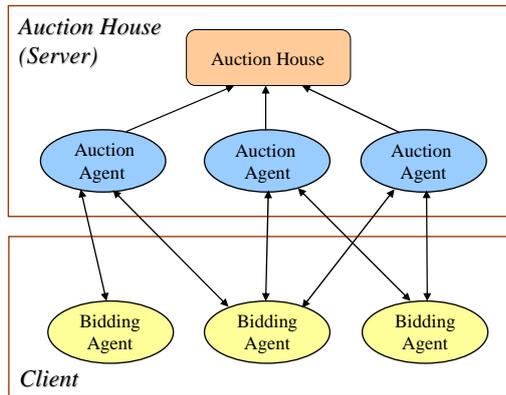
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Online Auctions

- Different types of auctions
 - Increase-price auction (English auction)
 - Decrease-price auction (Dutch auction)
 - Second-price sealed-bid auction (Vickrey auction)
- English auction has become the most popular one in online auction houses (e.g., eBay).
- However, it is time-consuming for a human user to search and place bids on an auctioned item.
- There is a pressing need to introduce agent technology into online auction systems.



Agent-Based Online Auctions



- It consists of an auction house and a number of clients.
- It is designed as a multi-agent system.
- The auction house is managed by auction house administrator.
- Agents at the client side work on behalf of human users.



Bidding agents can place bids on behalf of human users, and they are typically running on different machines.

Specification of Bidding Strategies

- Provide an interface for human users to specify bidding strategies
 - Complex and flexible bidding strategies
 - Easy-to-use interface
- Develop a formal model for bidding strategy specification
 - Separation of concerns: a layered approach
 - Adopt a visual modeling language, e.g., UML activity diagram
 - Convert it into a formal rule-based bidding model
- Advantages of allowing customized bidding strategies
 - Simulate human bidding behaviors more precisely
 - Support generation of real-time auction data for analysis and testing purpose

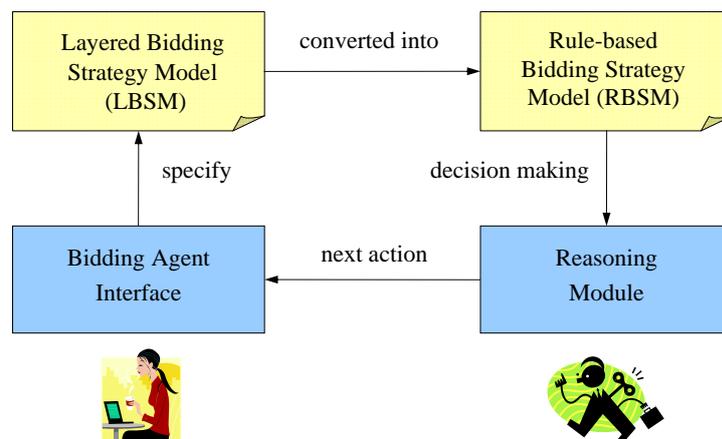


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Bidding Agent Architecture

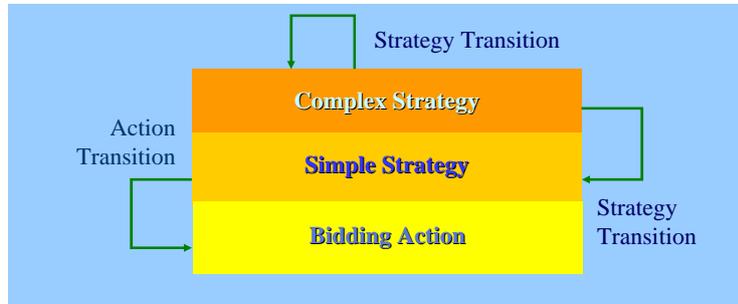


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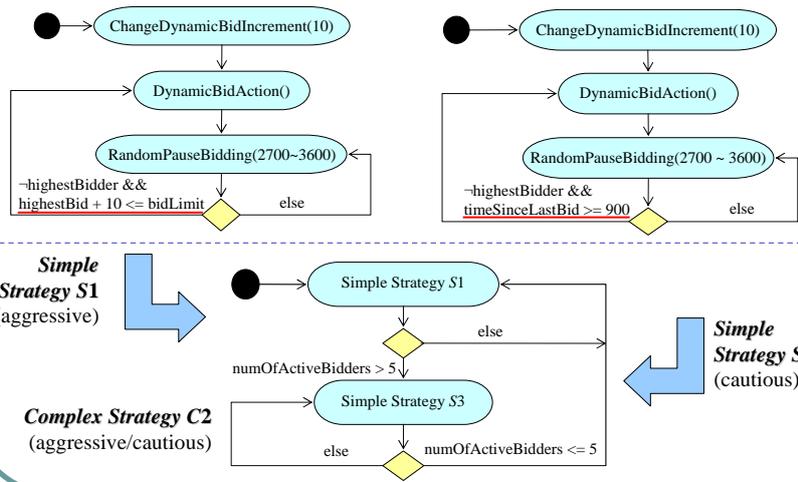
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Layered Bidding Strategy Model (LBSM)



- Examples of Bidding Actions**
- DynamicBidAction
 - ChangeDynamicBidIncrement
 - RandomPauseBidding(2700~3600)

Simple Strategy and Complex Strategy



Rule-Based Bidding Strategy Model

```
<production rule> ::= <strategy rule> | <action rule> |  
  <initial strategy rule> | <initial action rule>  
<strategy rule> ::= <s-domain> <bidding strategy> <condition> ->  
  <bidding strategy>  
<s-domain> ::= <s-domain>.<complex strategy> | <complex strategy>  
<bidding strategy> ::= <simple strategy> | <complex strategy>  
<condition> ::= <compound condition> | <arithmetic condition> |  
  <comparison condition> | <boolean condition>  
<action rule> ::= <a-domain> <action> <condition> -> <action>  
<a-domain> ::= <s-domain>.<simple strategy> | <simple strategy>  
<action> ::= <basic bid> | <change bid limit> | <change dynamic  
  bid increment> | <dynamic bid> | ... | <pause> | <stop>  
<initial strategy rule> ::= <complex strategy> -> <initial  
  strategy>  
<initial action rule> ::= <simple strategy> -> <initial action>
```

Definition of bidding strategy language (BSL) in Backus-Naur Form (BNF)

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Algorithm: Model Conversion

```
function convertToRuleBasedStrategyModel (LBSM lbsm)  
  if lbsm is a complex strategy  
    add a new initial strategy rule:  
      lbsm → lbsm.initialStrategy  
    for each StrategyTransition st in lbsm  
      set up s-domain according to the strategy hierarchy  
      add a new strategy rule: s-domain, st.startStrategy,  
        st.condition → st.endStrategy  
    end  
    for each strategy s in lbsm  
      convertToRuleBasedStrategyModel (s)  
    end  
  else if lbsm is a simple strategy /* base case */  
    add a new initial action rule: lbsm → lbsm.initialAction  
    for each ActionTransition at in lbsm  
      set up a-domain according to the strategy hierarchy  
      add a new action rule: a-domain, at.startAction,  
        at.condition → at.endAction  
    end  
  end function
```

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Algorithm: Reasoning Engine

```

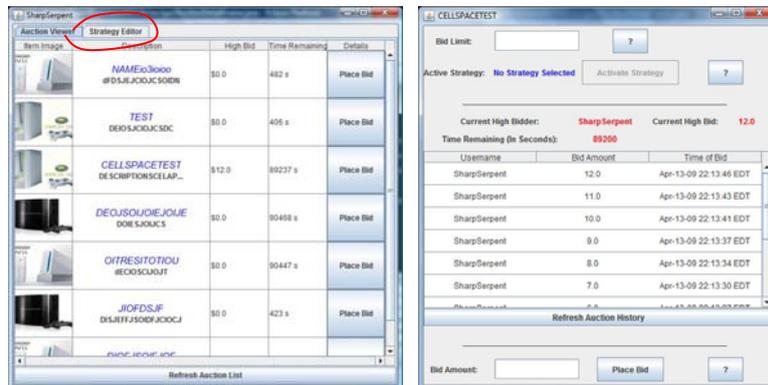
function Action findNextAction
  (Domain domain, Action currentAction)
  if currentAction == null
    if domain is a ComplexStrategy
      Search for initial strategy rule isr for
      domain that leads to initial strategy is
      return findNextAction (is, null)
    else if domain is a SimpleStrategy
      Search for initial action rule iar for
      domain that leads to initial action ia
      return ia
    else if currentAction != null
      Remove and process the first element
      fe of domain, and let the remaining
      domain be r-domain
      if fe is a ComplexStrategy
        /* strategy transition */
        Retrieve all strategy rules for the first
        element of r-domain- and store them
        in a list
      while the list is not empty
        Remove and process strategy rule sr at list
        head
        if the condition for sr is true
          Let s be the conclusion part of sr
          return findNextAction (s, null)
        return findNextAction (r-domain, currentAction)
      else if fe is a SimpleStrategy
        /* action transition */
        Retrieve all action rules for the currentAction
        and store them in a list
        while the list is not empty
          Remove and process action rule ar at list
          head
          if the condition for ar is true
            return the conclusion part of ar
          return currentAction
      end function
  
```

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Bidding Agent Interface

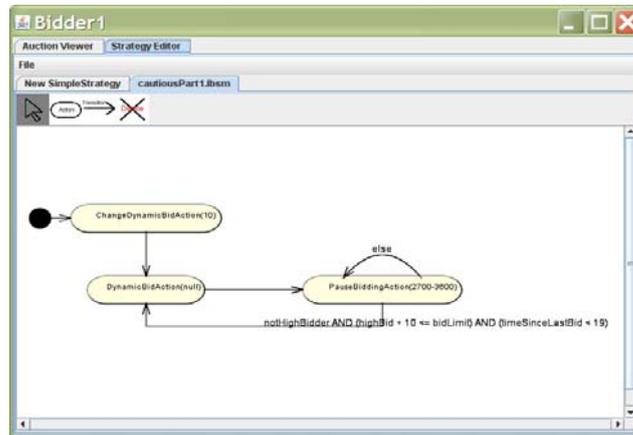


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Designing Bidding Strategies



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Case Study

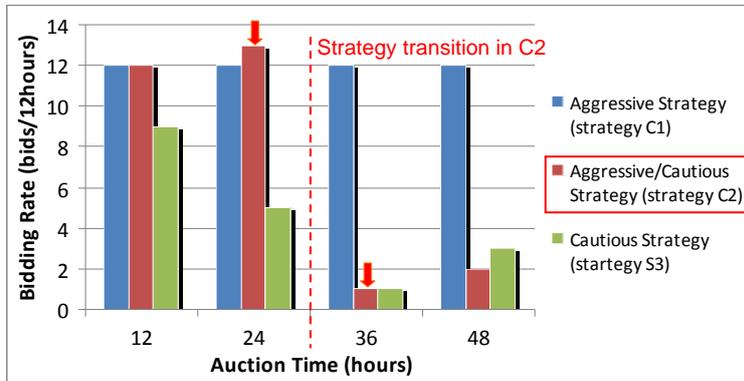
- A fictitious auction: an item with an estimated auction price of \$1000.
- Six bidding agents: Bidder 1 to Bidder 6
 - Bidder 1-5 are active through the whole auction
 - Bidder 6 joins at the middle of the auction
- Run the auction three times with Bidder 2-6 using a *normal* strategy, but Bidder 1 using one of the three different bidding strategies:
 - Aggressive strategy C1 – composed of simple strategy S1 and S2.
 - Aggressive/Cautious strategy C2 – composed of simple strategy S1 and S3.
 - Cautious strategy S3 – simple strategy.

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Experimental Result - 1



Bidding Rates with Different Strategies

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Experimental Result - 2



Bidding Prices with Different Strategies

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Conclusions and Future Work

- Provided a model-based specification approach for complex bidding strategies.
- Demonstrated how bidding agents can automatically place bids according to bidding strategies.
- Used a case study to show how our prototype can support analysis of agent-based online auctions.
- For our future work, we plan to improve the GUI for visual specification of complex bidding strategies.
- Use agent-based online auction system as a test bed for skill detection.
- Implement a trustworthy agent-based online auction system.

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Questions?

The slides for this talk can be downloaded from
<http://www.cis.umassd.edu/~hxu/Projects/ATM>

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