Distributed Reinforcement Learning for Electricity Market Bidding with Spark

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Speaker Bios

- Vishnu is a Senior Data Scientist with experience in Reinforcement Learning, Supervised Machine Learning, Stochastic Optimization, and Statistical Analysis. He was a faculty member at the University of Wisconsin-Milwaukee before joining Impetus.
- Vijay is the Director for BigData Labs at Impetus, with experience in cloud, grid, peer-to-peer computing, and machine learning for Big-Data. He is the author of the recently published book "Big Data Analytics Beyond Hadoop."

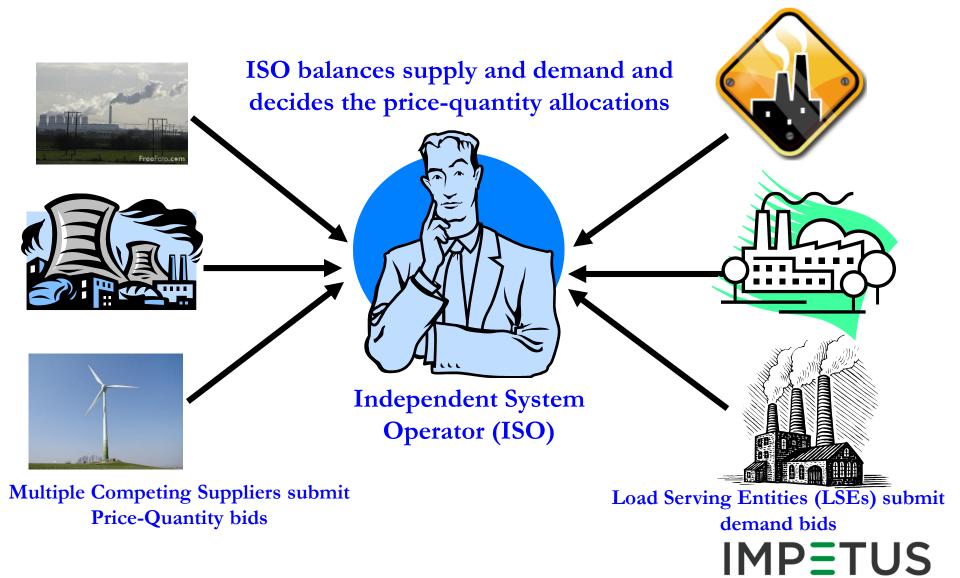


Agenda

- What is an electricity market?
- Why does it need to be modeled?
- Are there other markets that need/or could use such modeling?
- Why reinforcement learning?
 - Background and basics
- Why Spark? What other options exist?
- Solution strategy over spark
- Ongoing and future work



What is an Electricity Market



Innovation Architected

Electricity Markets in North America





http://www.caiso.com/about/Pages/OurBusiness/ UnderstandingtheISO/Opening-access.aspx



Why does it need to be modeled?

- To understand long-term behavior of market
- Equilibrium behavior
- Assessment of market power
- Investigate policy prescriptions



Modeling Approach

- ☐ Bid each day (each hour) for supplying power the following day
- ☐ Stochastic processes underlying the day ahead energy market operations are modeled as a Markov chain
- ☐ Decisions embedded on a Markov chain are modeled as Markov decision processes
- ☐ Multiple generators competing in a market is modeled as a competitive Markov decision process (a.k.a., stochastic game)



CMDP Model

```
\Box Let B = Set of Supply/Load Buses in the Network
    M = \# of loads
     N = \# of suppliers
        X^{t} = (q^{t}, p^{t}), is the system state at any day 't'
where,
       q^t = (q_1^t, q_2^t, q_3^t, \dots, q_{|B|}^t)-forecast load vector
       p^t = (p_1^t, p_2^t, p_3^t, \dots p_{|B|}^t)- anticipated bus
                                                 price
                                                         vector
       D^{t} = (D_{i}^{t} : I \in \{1,2,3,....N\})
where,
\Box D_i^t = I^{th} generator's bid
```



CMDP Model

- $\square X = \{X^t : t \in N\}$ System State process
- $\square D = \{D^t : t \in N\}$ Decision process
- \Box The joint process (X, D) is a Competitive Markov Decision Process
- Average reward criterion and non-zero sum nature of the game make solution more challenging
- ☐ Hence, simulation based optimization called Reinforcement Learning (RL)



RL Basics

- The theory of RL is founded on two important principles:
- Bellman's equation and the theory of stochastic approximation.
- Any reinforcement learning model contains four basic elements:
 - System environment (simulation model)
 - 2. Learning agents (market participants)
 - 3. Set of actions for each agent (action spaces)
 - 4. System response (participant rewards)

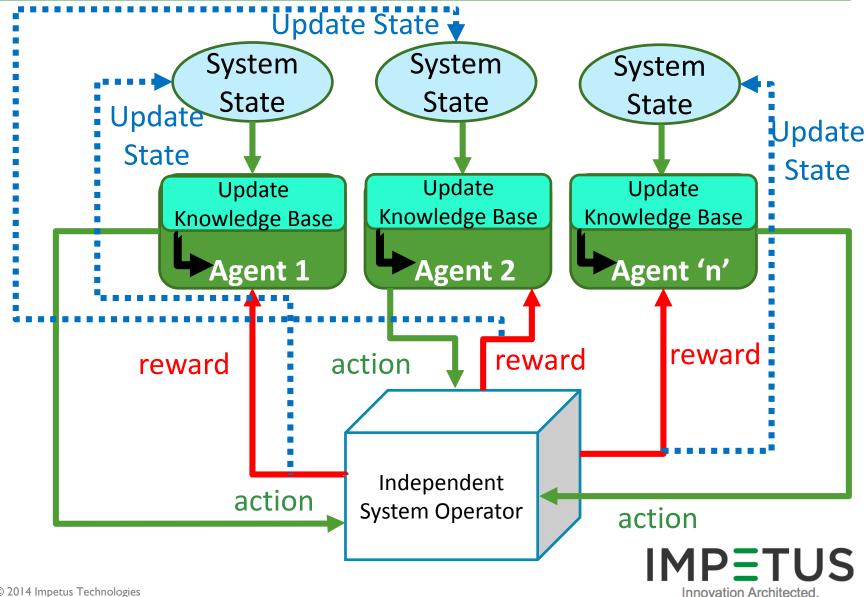


Are there other applications?

- Generation capacity expansion planning
- Carbon allowance markets



Reinforcement Learning: A Schematic

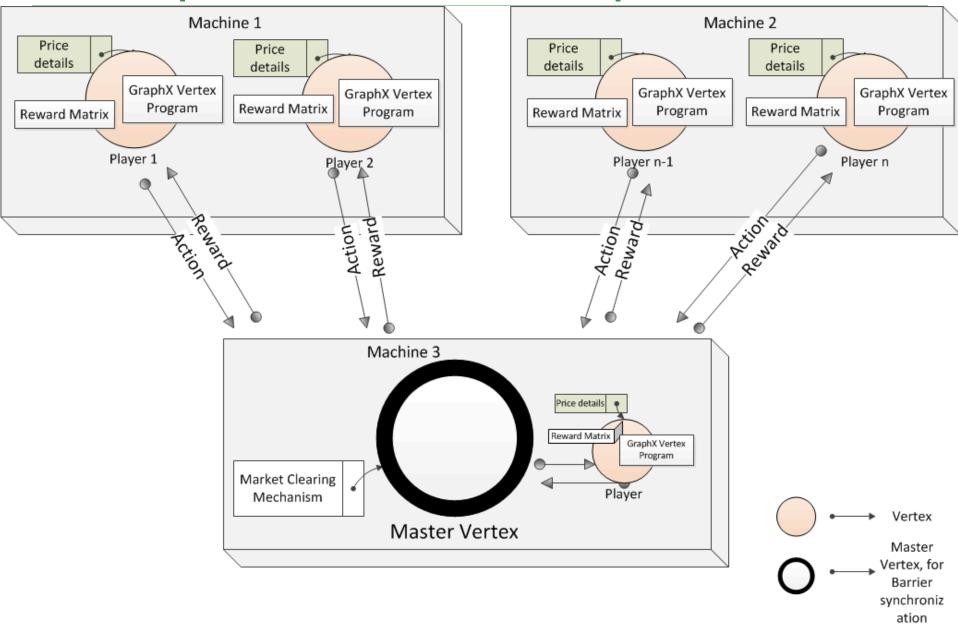


Why Spark?

- Hadoop Map-Reduce is not well suited for iterative machine learning algorithms
- Twister/HaLoop/Apache Hama
 - Fault-tolerance is questionable,
 - Enterprise readiness is not clear
- GraphLab
 - Can only take distributed snapshots no automatic recovery.
- Possible application to the spot market realtime electricity bidding.



RL Implementation over Spark



Work-in-progress

- Giraph like APIs for GraphX
 - Allow user to specify master code VS slave code
 - Explicit barrier synchronization
- Reinforcement learning over Spark
 - Can be used by independent system operator or by individual generators
 - Can be used in other competitive bidding-based markets such as carbon allowance markets, which is a multi-million dollar industry in U.S. and Europe.
- RL over Hadoop/Stratosphere
 - Comparison with Spark/GraphX



Thank you!

