



# Toward a Causal Model for Automatic Game Balancing

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## Overview

- Goals of Research
- Needs of Automatic Game Balancing
- Toward a Causal Model
  - Defining a Causal Model
  - Building the Model
  - Validating the Model: Player Study
- Future Work: Applying the Model

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## Goals of Research

- Make game balancing more scientific
  - Justify changes made to game
  - Model relationships between game factors
  - Create function to optimize
- Build automatic game balancing system
  - Analyze game session
  - Adjust game accordingly
  - Increase fun by balancing game

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## Needs of Automatic Game Balancing System

- Quantitative
  - Need explicit objective function
  - No way to get data from user post release
- Utilize accessible data from system
  - Available data: health, accuracy, etc.
  - Bio-sensors impractical
- Right level of abstraction
  - High level not useful
  - Need to encompass all games
- Changes made justified by data

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## Prior Work

- Flow (Csikszentmihalyi)
- EVE (Burns)
- MDA (Hunicke, LeBlanc, Zubek)
- GameFlow (Sweetser and Wyeth)
- Predator/Prey Heuristic (Yannakakis and Hallam)

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## Overview

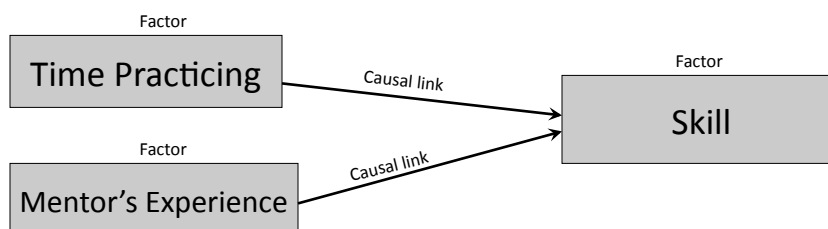
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## Causal Model

- Abstract model using cause and effect
- **Factors**: Represent variables in model
- **Causal links**: express cause and effect relationship between factors
- Clearly model interaction of game factors



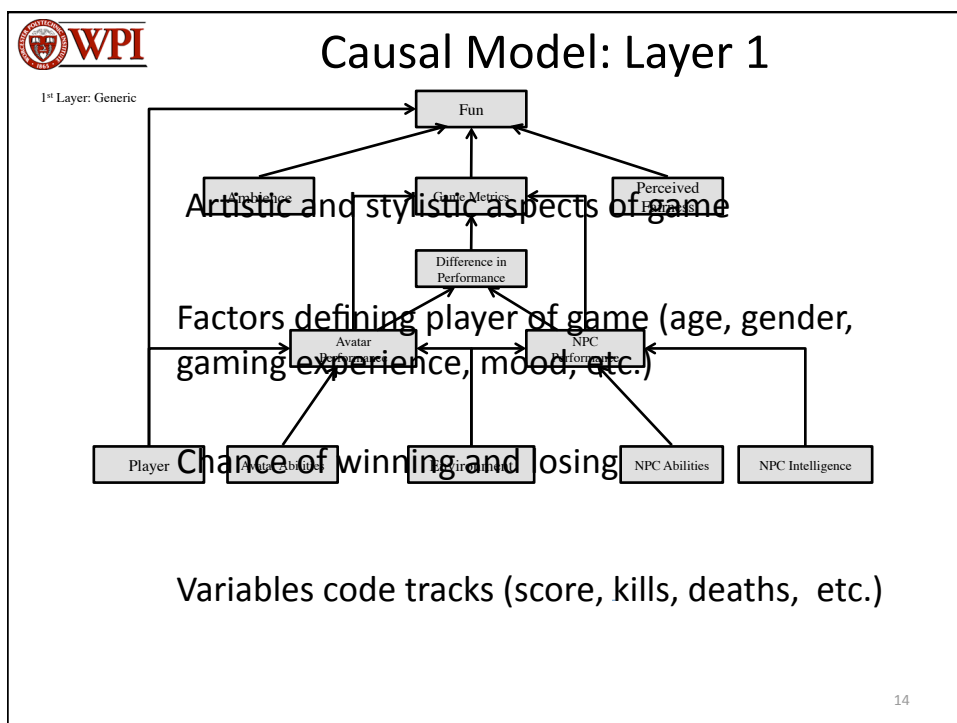
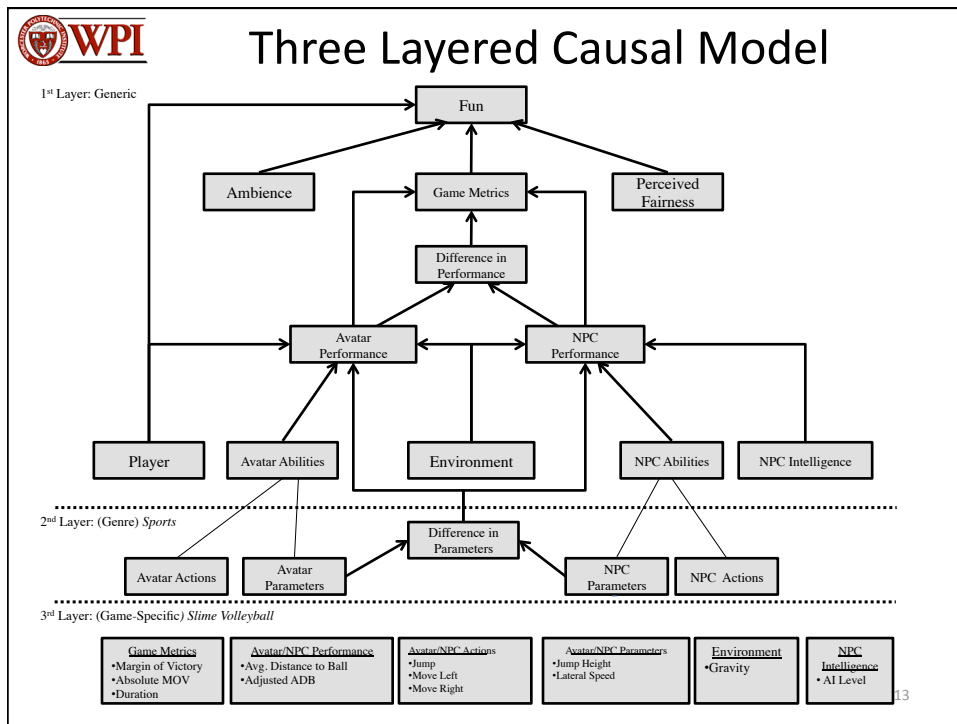
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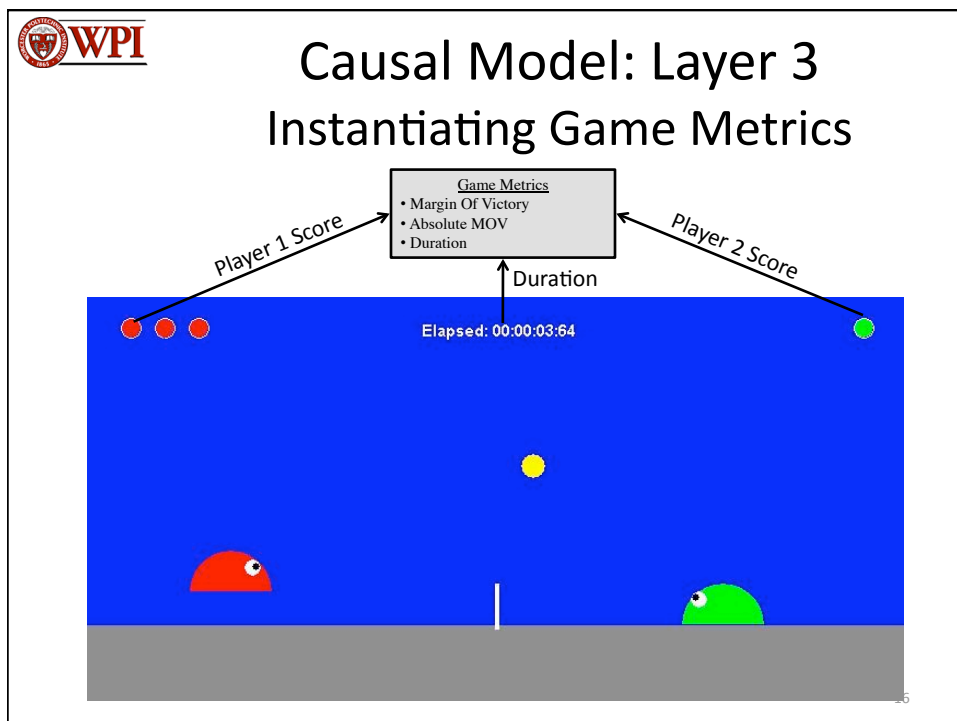
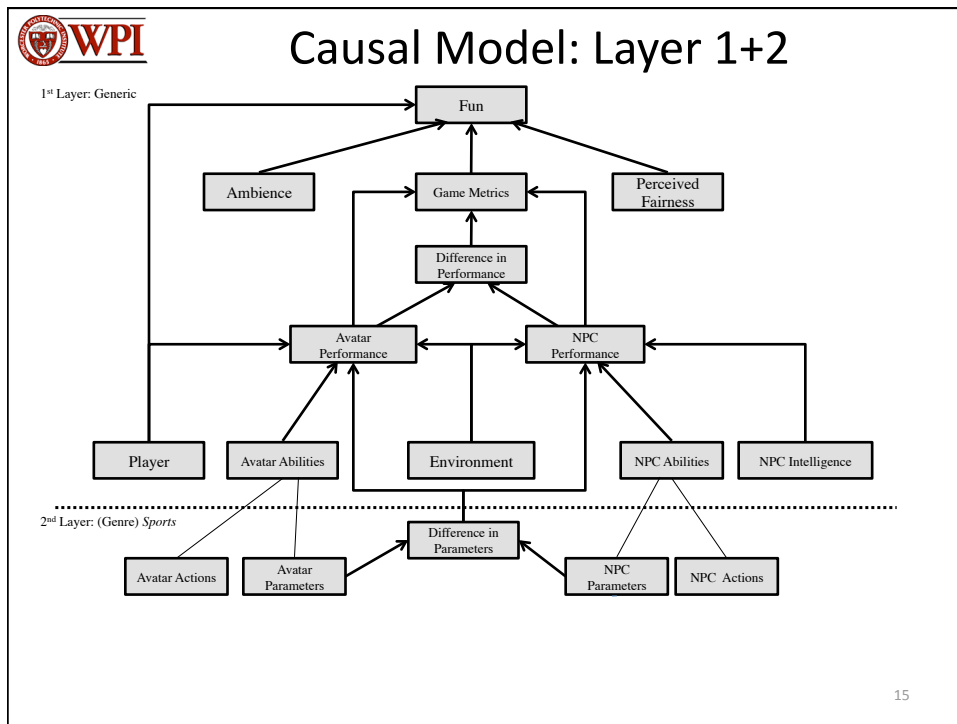


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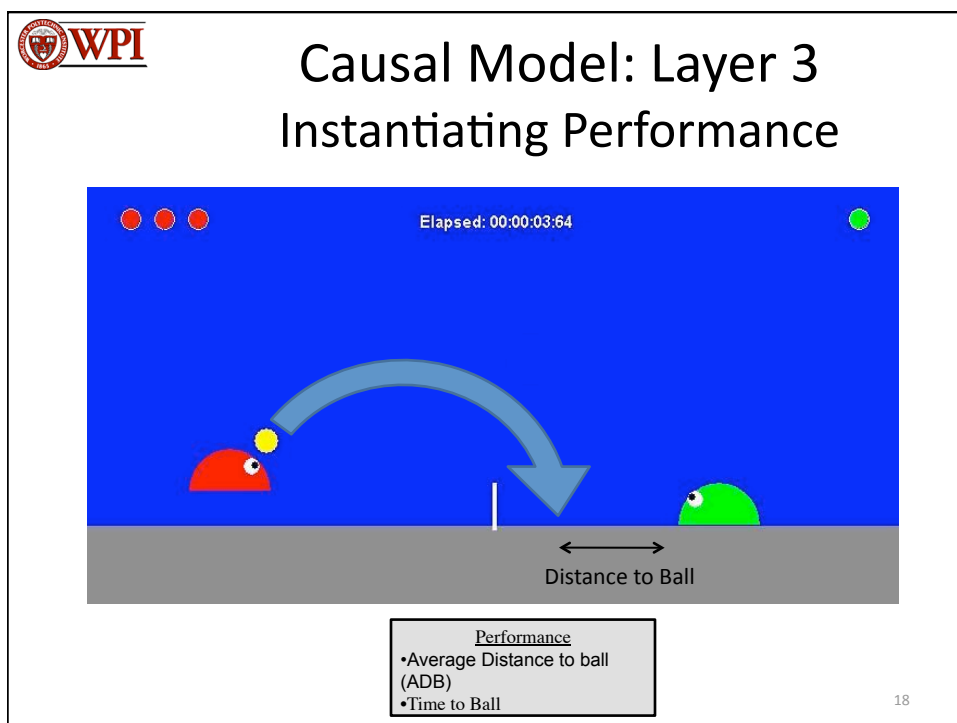
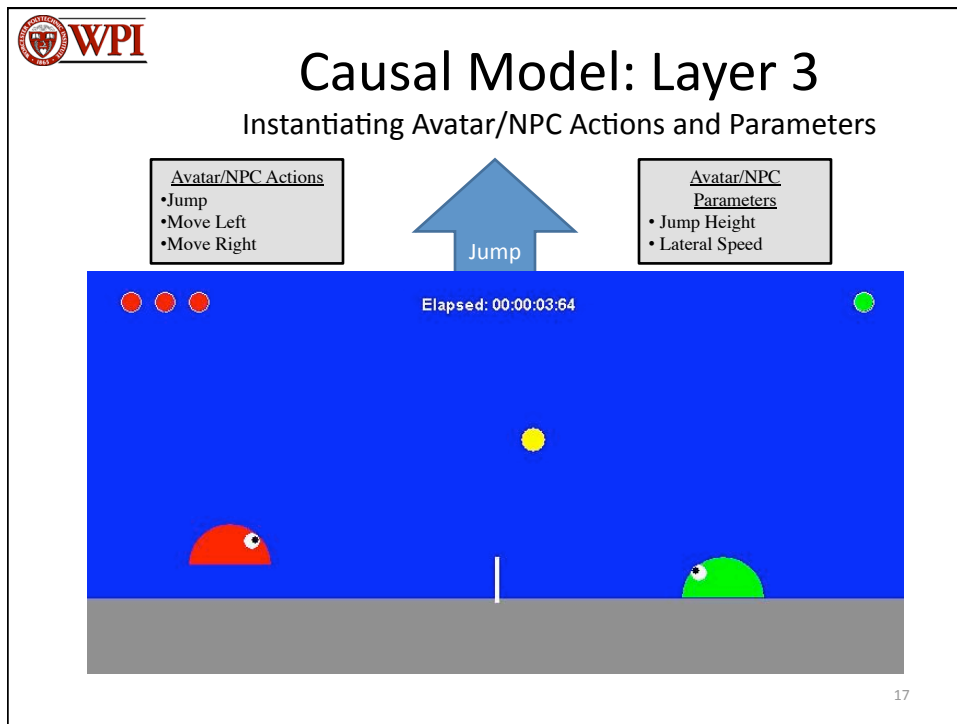
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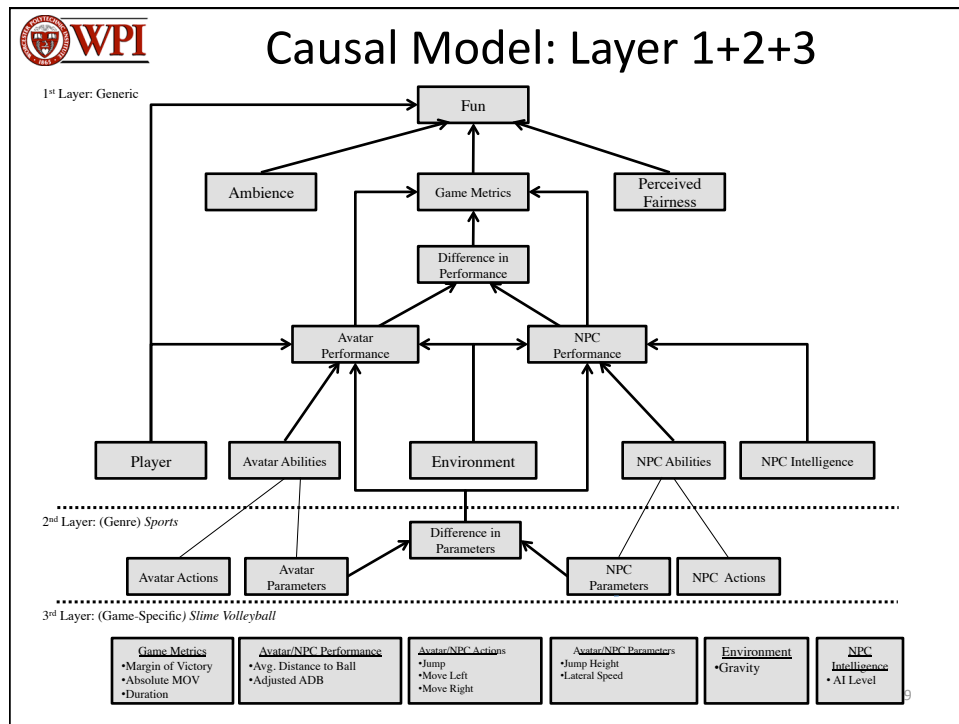
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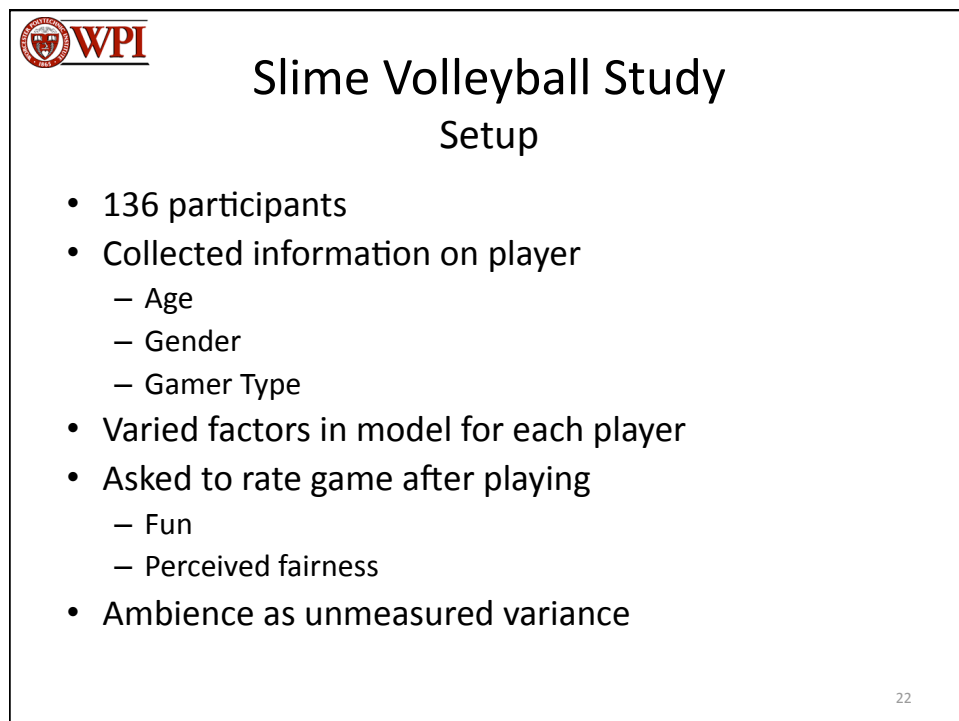
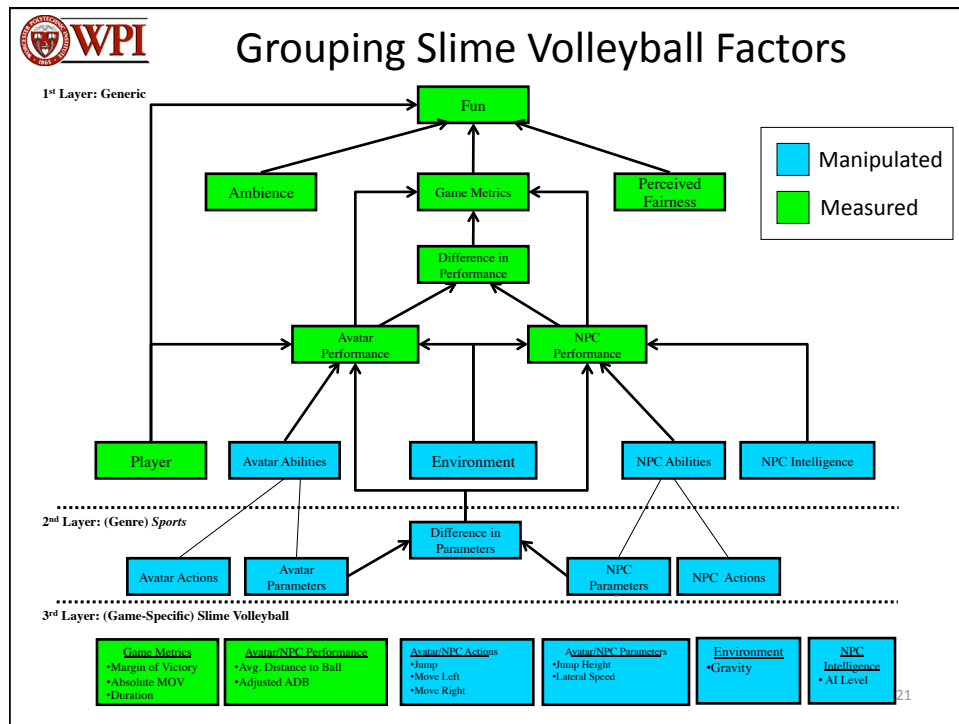




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## Slime Volleyball Study

### Pre-game Survey

#### Age

- Under 20
- 20 to 25
- 26 to 30
- 31 plus

#### Gender

- Male
- Female

#### Gamer Type

- Casual
- Normal
- Hardcore
- Unknown

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## Slime Volleyball Study

### Post-game Survey

- How much fun was this match? (0-9)
  - 0: No Fun
  - 9: Very Fun
- How fair was this match? (0-9 where 0 means computer had advantage, 5 was an even match, and 9 means you had the advantage)
  - Perceived Fairness
  - Binned Fairness
  - Absolute Fairness

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## Slime Volleyball Study Analysis

- Justify causal links with correlations
  - Highly correlated ( $r > .32$ )
  - Remove if not justified
- Test causal links not in model
  - Correlation
  - Add if justified
- Determine how much variance we account for

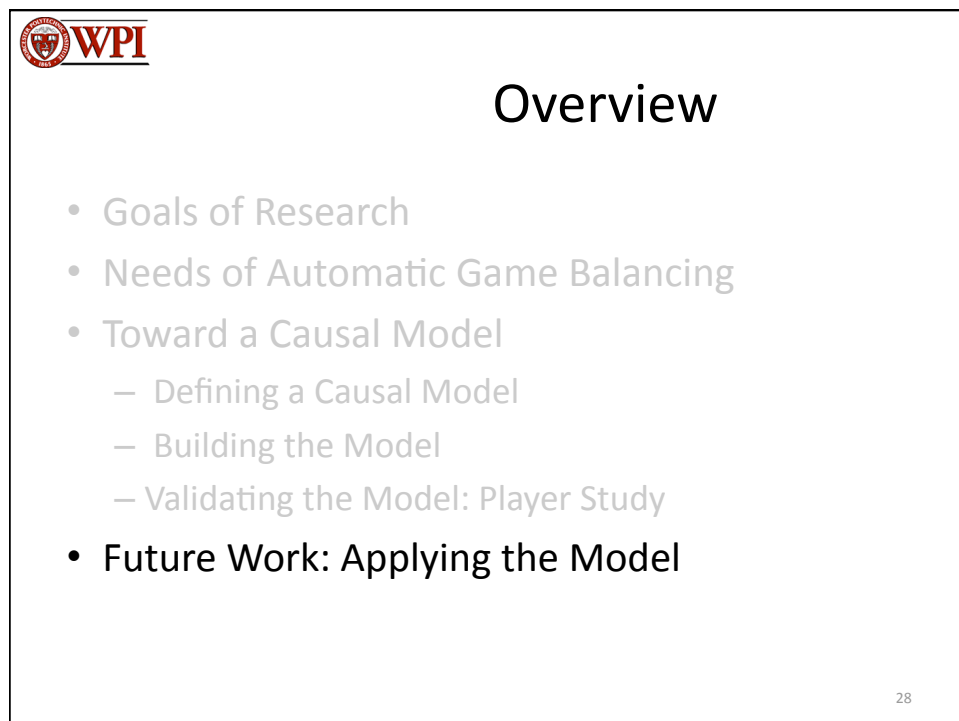
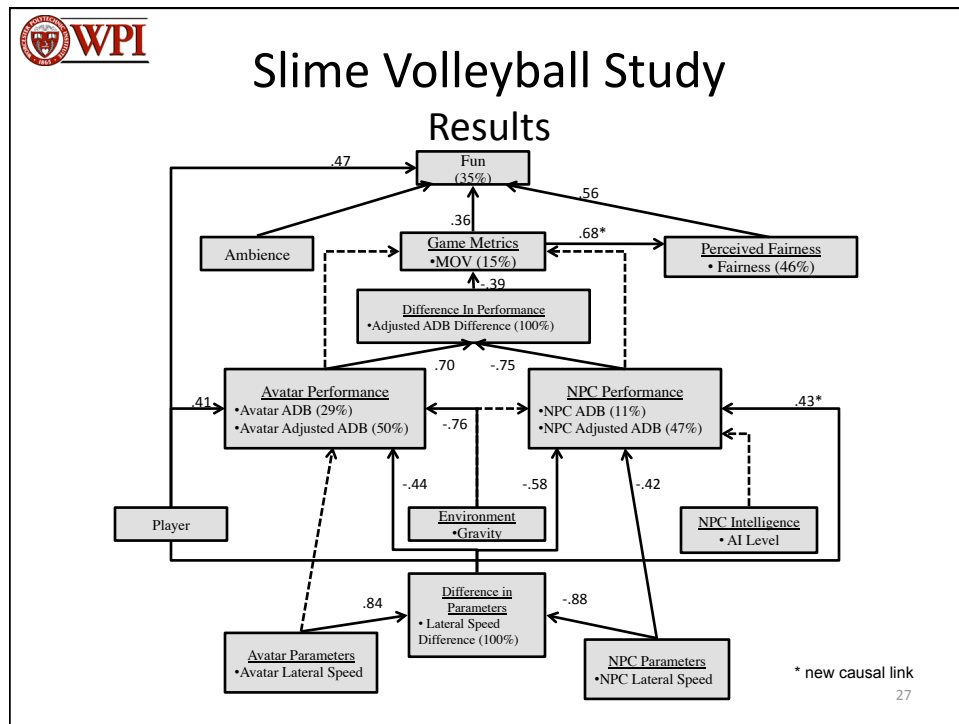
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## Slime Volleyball Study Using Correlations

- Correlation does not imply causality!!!
  - Correct
  - We are not using correlation to create causal links
- Building causal model from prior knowledge
  - Academic experience
  - Research
  - Game playing experience
- Using correlations to support proposed causality

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## Future Work: Applying the Model

- What in the model will be altered?
  - Only alter values in nodes
  - Alter values and structure of model
- How should the model be altered?
  - Linear Regression
  - Reinforcement Learning
  - Other
- What game should be used to test automatic game balancing system?
  - Utilizes all nodes
  - Expand to other genre
  - Able to change node values easily
- Finished in May 2010 for Master Thesis

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## Ongoing Study

- Instantiating new causal model
- Participants needed
  - <http://moffett-research.genbook.com/>
    - A25 Fuller Laboratory
    - Play on local machine
  - <http://users.wpi.edu/~jeffmoffett>
    - Download game
    - Play from home
    - Email results

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# Questions?



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