

Flatland Visualization of A Decision Support Tool Architecture

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The Team

- **New Mexico State University**- Jim Cowie, Chris Fields, Hung Nguyen , Bill Ogden, Ram Prasad
- **Monterey Institute for International Studies**- Gary Ackerman, Markus Binder, Sundara Vadlamudi
- **University of New Mexico**- Frank Gilfeather, Thomas Caudell, Panaiotis, Tim Ross, Mahmoud Taha

The Problem

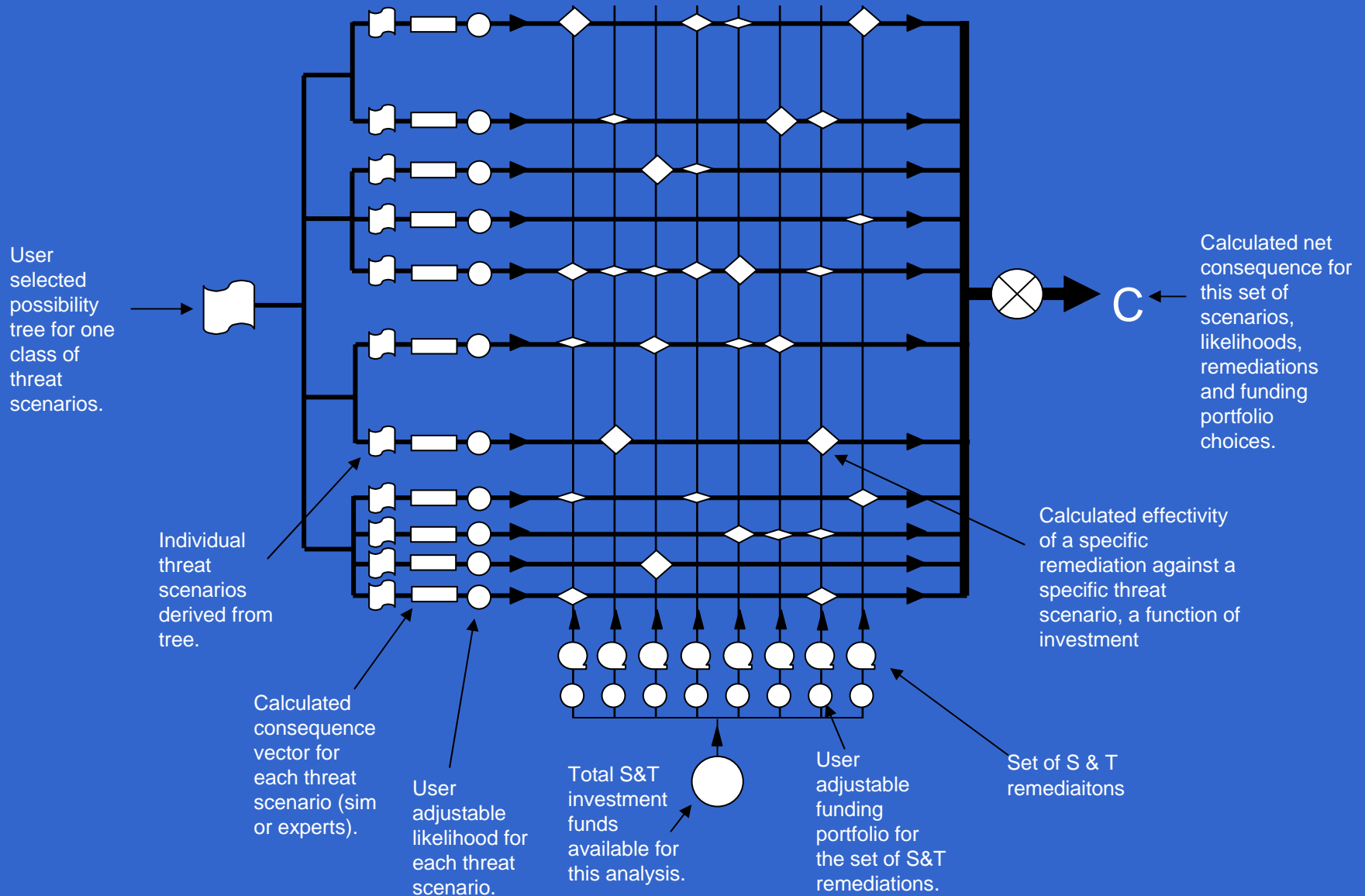
Allocation of Science & Technology (S&T) research funding to maximally reduce the threat and consequences of CB attacks on critical assets is complex and very hard to optimize globally.

Design Goals

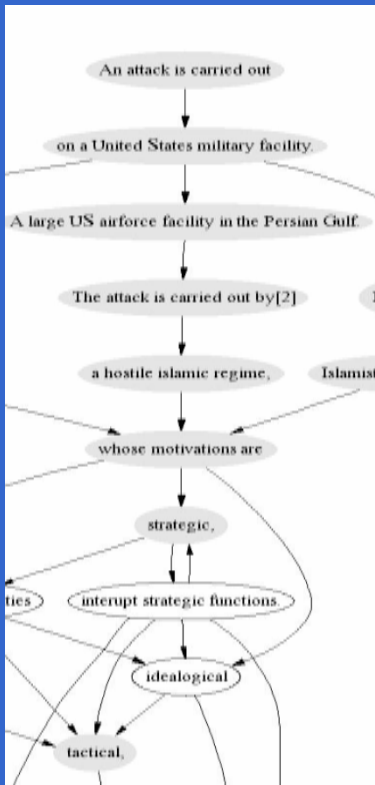
- Develop an analytic and algorithmic framework for threat consequence minimization.
- Create a feasible system architecture to evaluate modeling, analysis approaches, and user interactions within this framework.
- Enhance decision process transparency.

Initial Architecture

No Temporal Dynamics



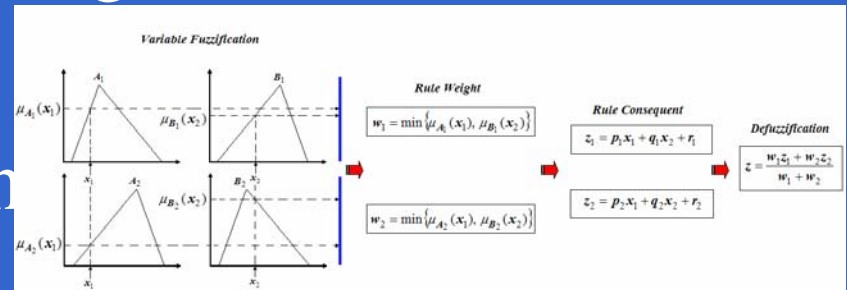
Specification of Scenarios



- Possibility Trees
- Spanning sets of scenarios
- Vectors of consequences per scenario
- Possible continuous scenario space
- Possible continuous consequence space

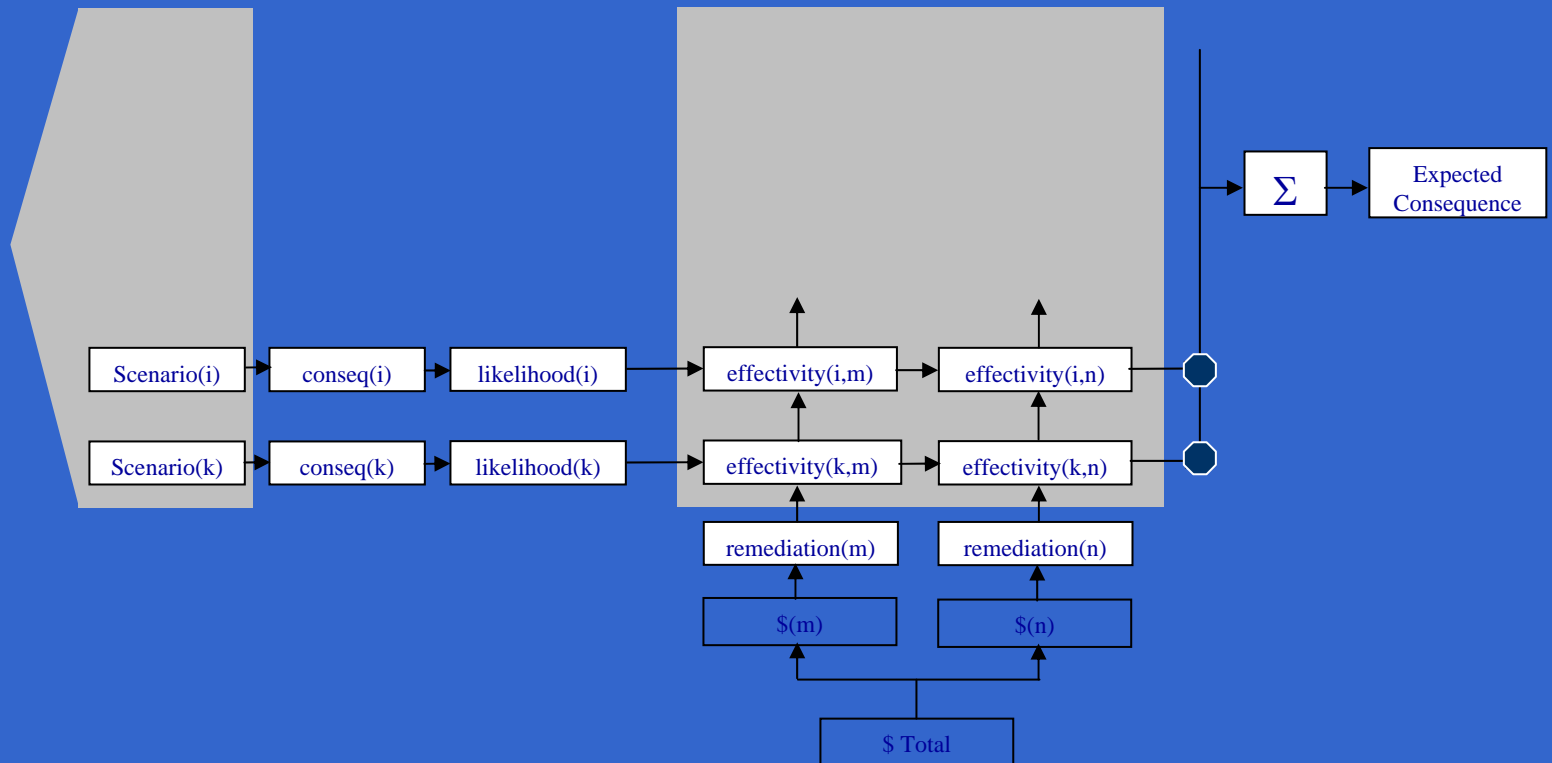
Creating Models of Consequences, Costs, and Effectiveness

- Relate remediation funding level to its effectiveness against a given scenario's consequences.
- Scientific Simulation
- Machine learning models
- Knowledge based systems



Details of Architecture

Consequence Flow Model



Mockup Mathematical Model

Assumes only one type of consequence per scenario.

$$\text{Expected Conseq} = \alpha \sum_{k=1}^{\text{NumScenarios}} \text{conseq}(k) * \text{likelihood}(k) * \prod_{m=1}^{\text{NumRemediations}} (1 - \text{effectivity}(k,m))$$

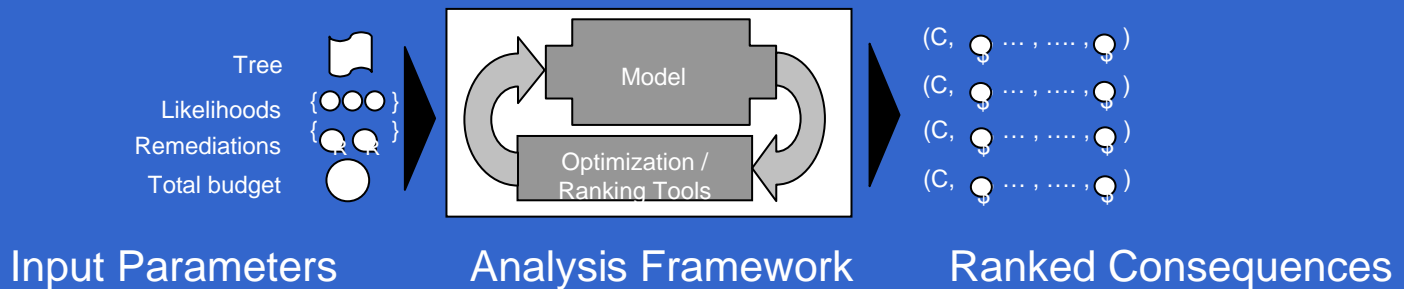
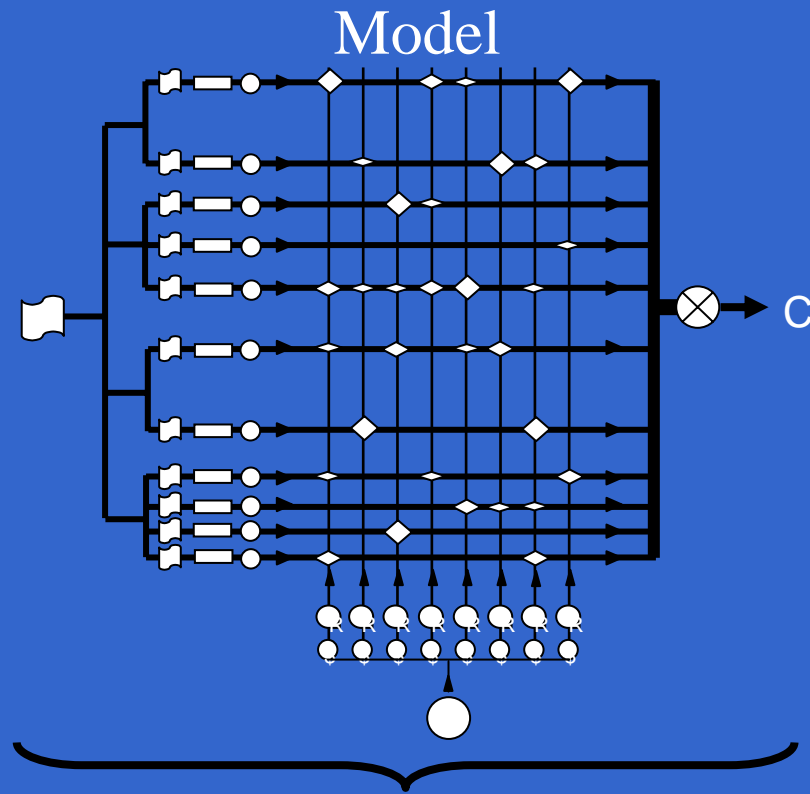
Where:

- * α is a normalization constant,
- * $\text{conseq}(k)$ is a real scalar,
- * $\text{likelihood}(k)$ is a probability (sums to 1),
- * $\$(m)$ is a real (sums to \$ Total = 1),
- * $\text{effectivity}(k,m) = \beta(k,m) * F(\$(m))$,
- * $\beta(k,m) = \text{rand}(0,1)$.

User Adjustables:

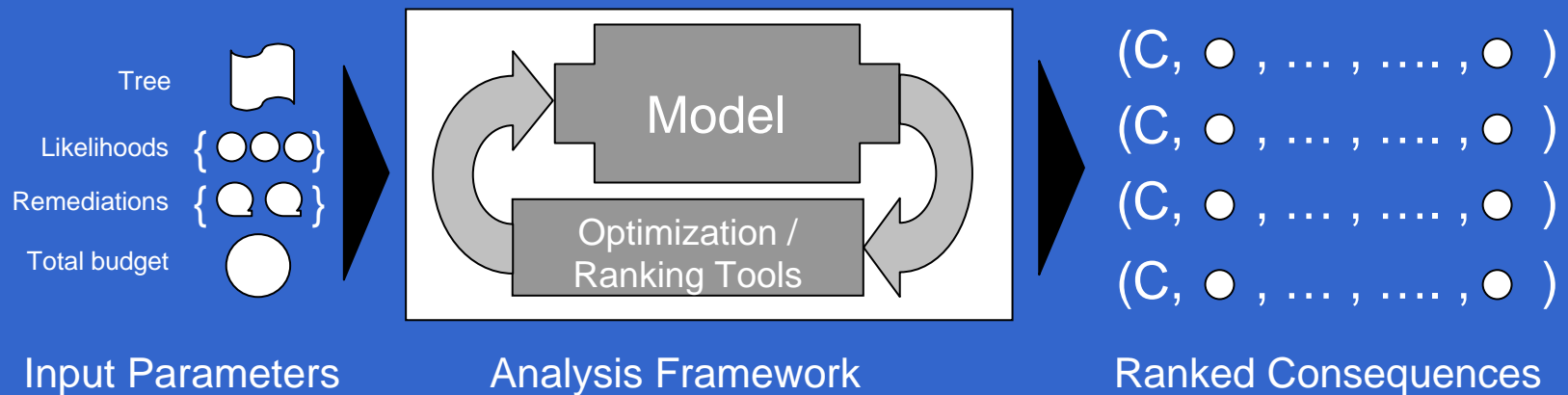
- a) { $\text{likelihood}(k)$, $k=1, \text{NumScenarios}$ }
- b) { $\$(m)$, $m=1, \text{NumRemediations}$ }

Optimization Loop



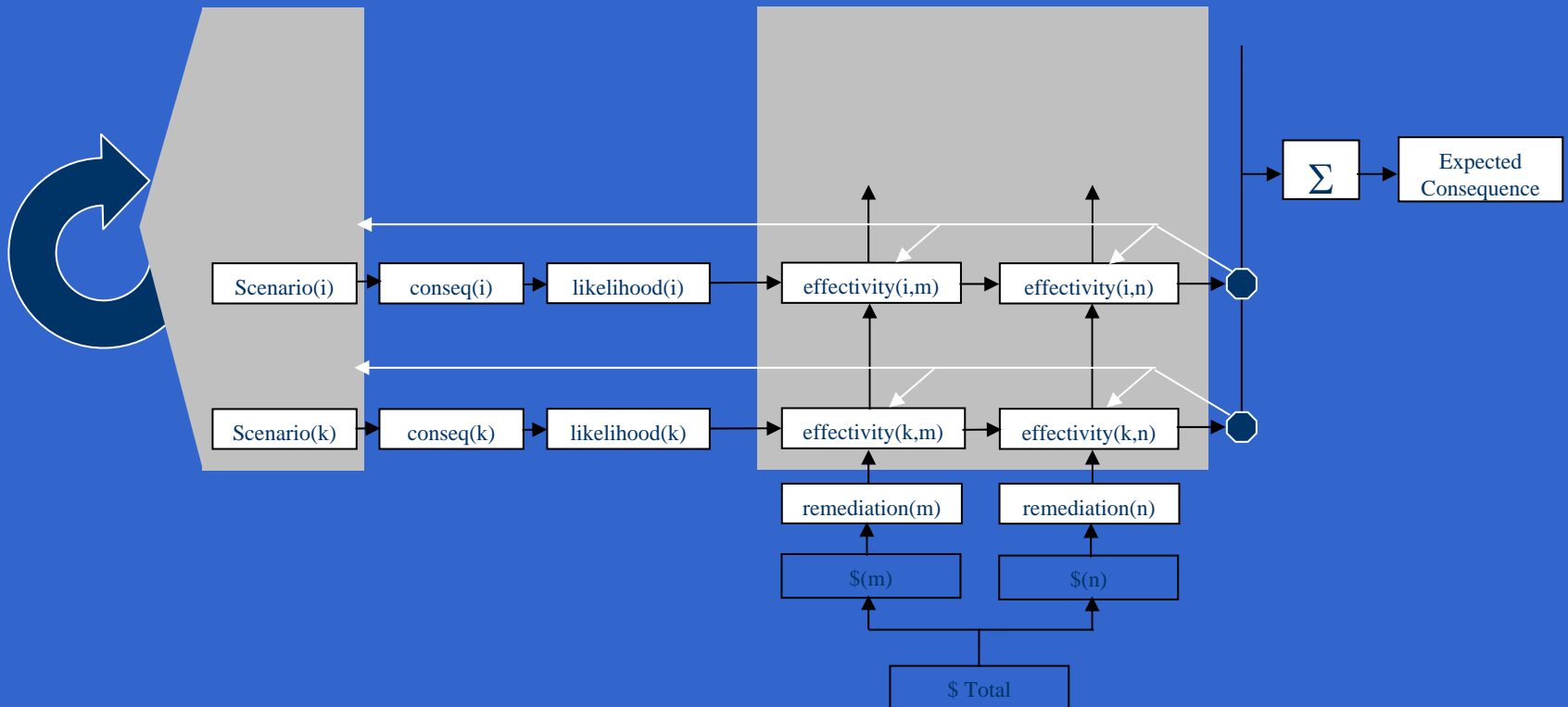
Optimization

Allocation of funds to minimize expected consequences



$$\text{Mockup: } \$_n(k+1) = \$_n(k) - \text{eta} * \text{gradient}(\text{Expected Consequences})$$

Temporal Dynamics



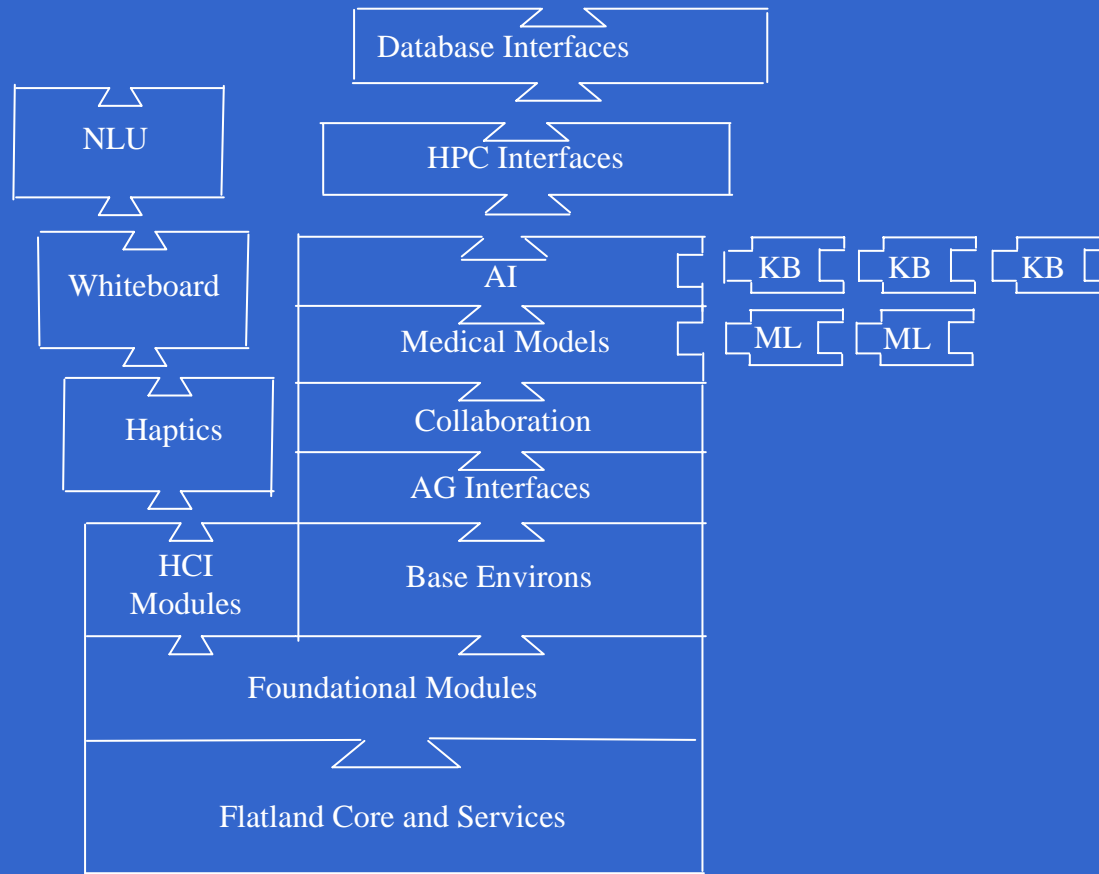
The visualization team

- Shan Xia - UNM, ECE
- Victor Vegara- UNM, ECE
- Panaiotis- UNM, ECE & Music
- Steve Smith- LANL
- Thomas Caudell- UNM, ECE & CS

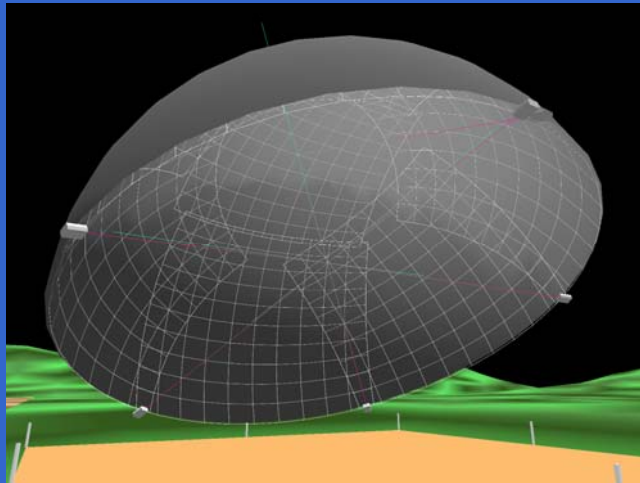
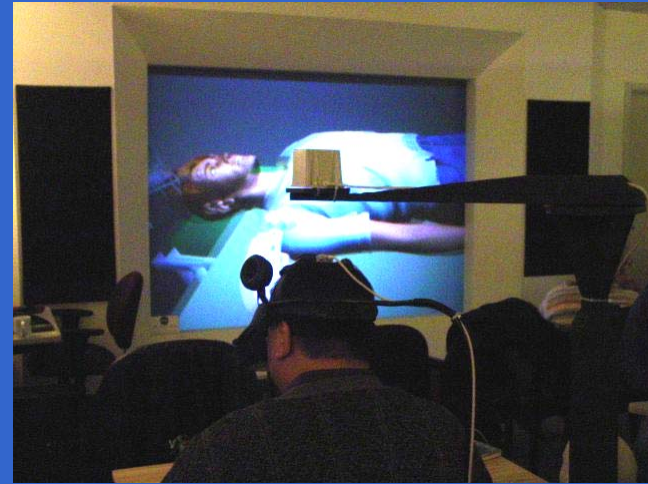
Features of visualization

- Goal: transparency into computational model and decision process.
- Consequence-flow metaphor
- Real-time user adjustable parameters
- User viewpoint control to manage complexity
- Drill-down for more details
- Animation of calculations and optimization
- Complementary sound representation of system states and dynamics
- Implemented in Flatland.

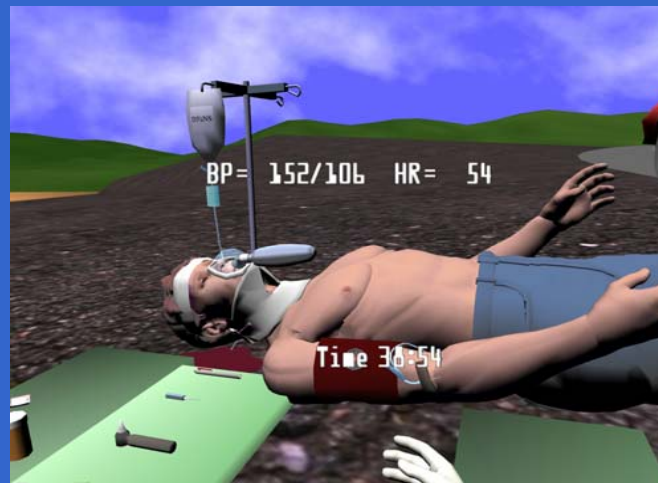
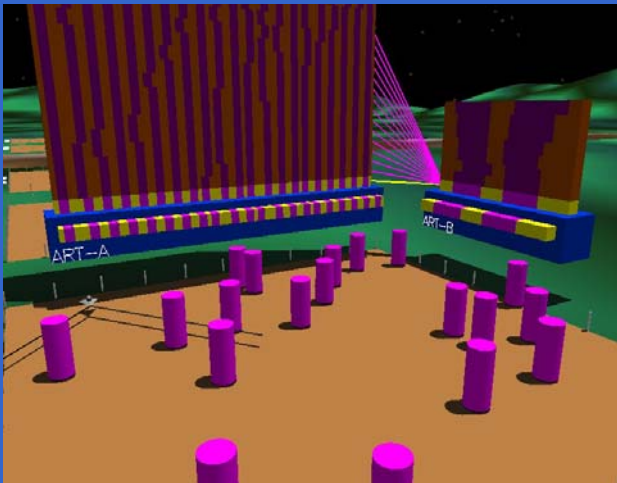
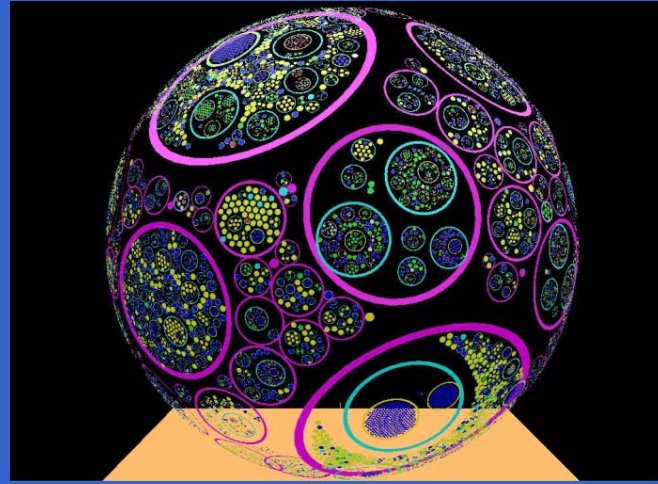
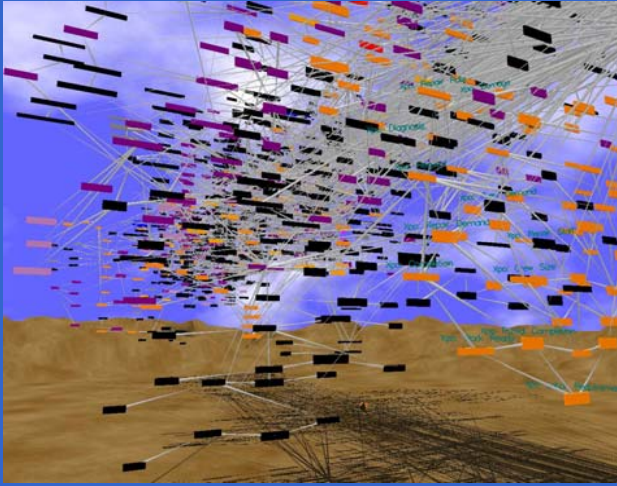
Flatland: modular applications



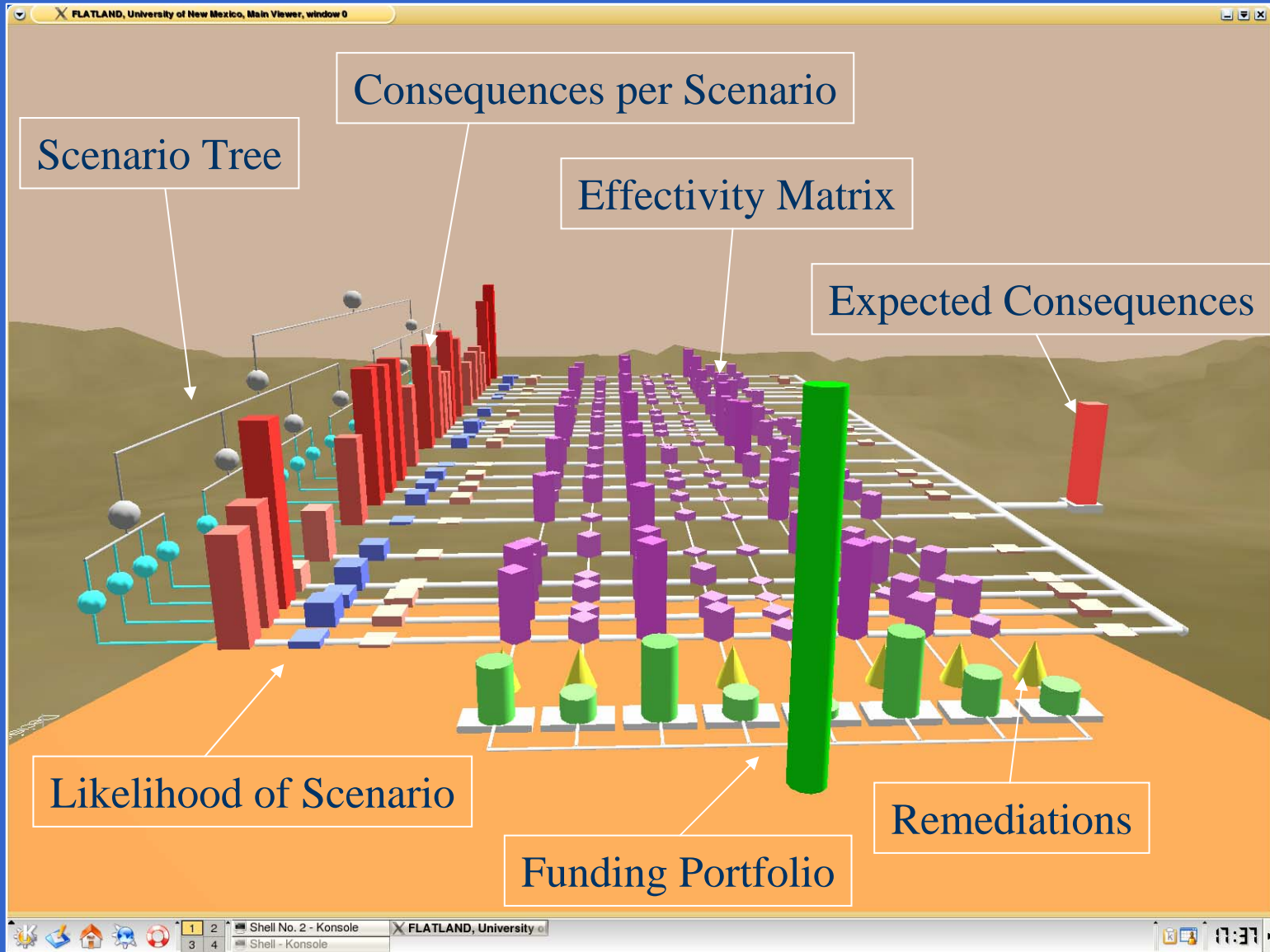
Displays: visual and sound



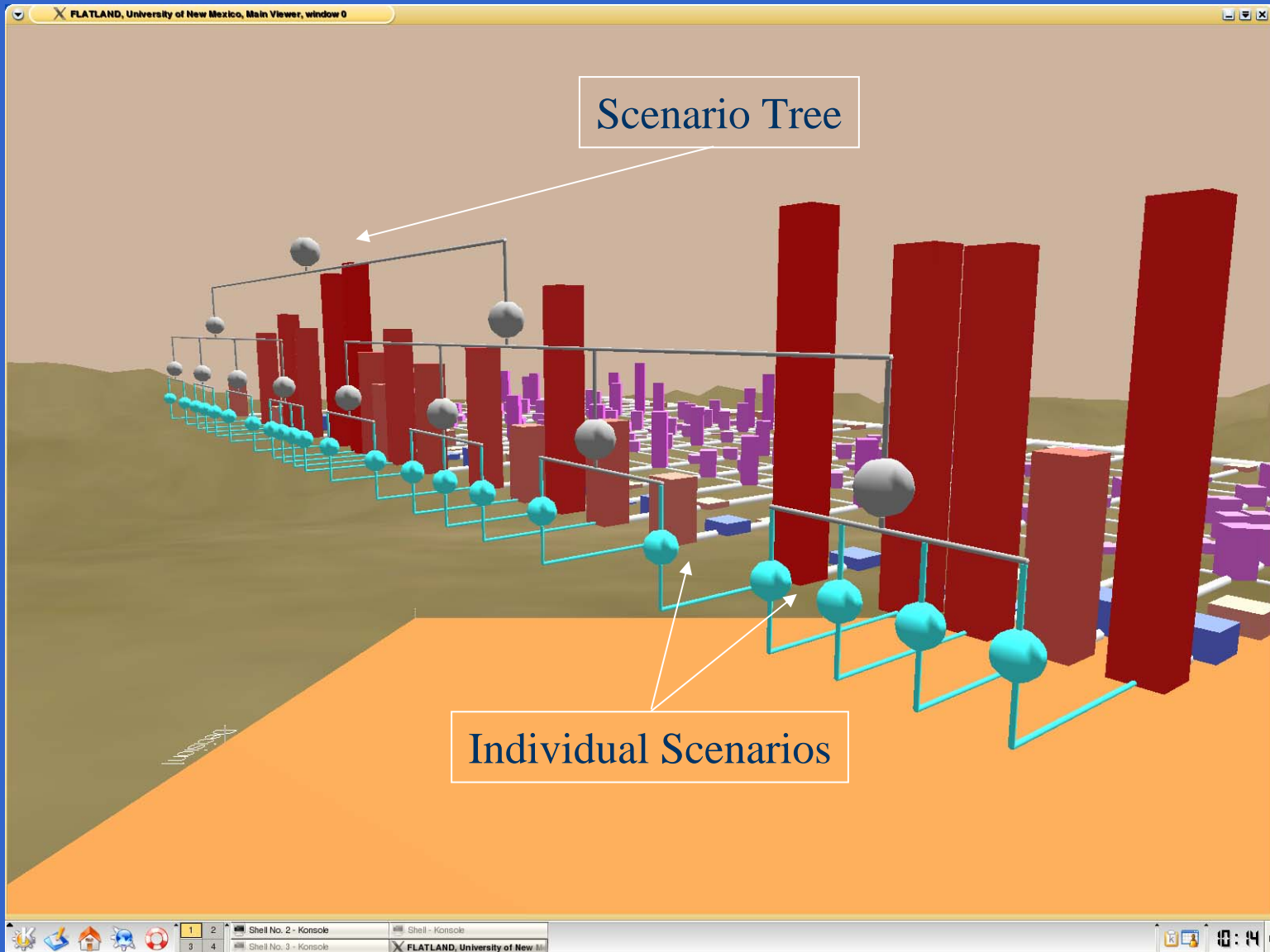
Widely applied



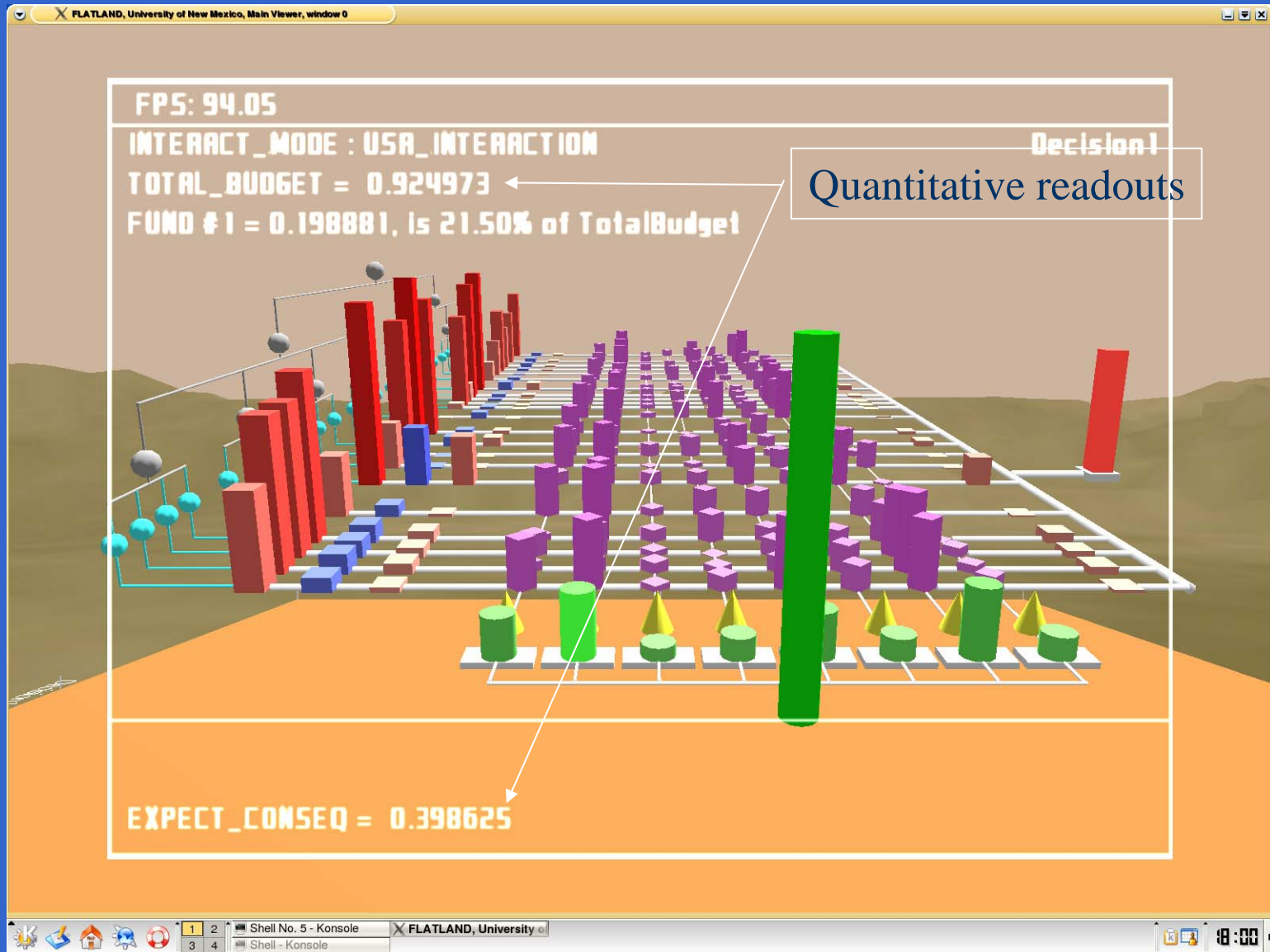
Visualization of Mockup System



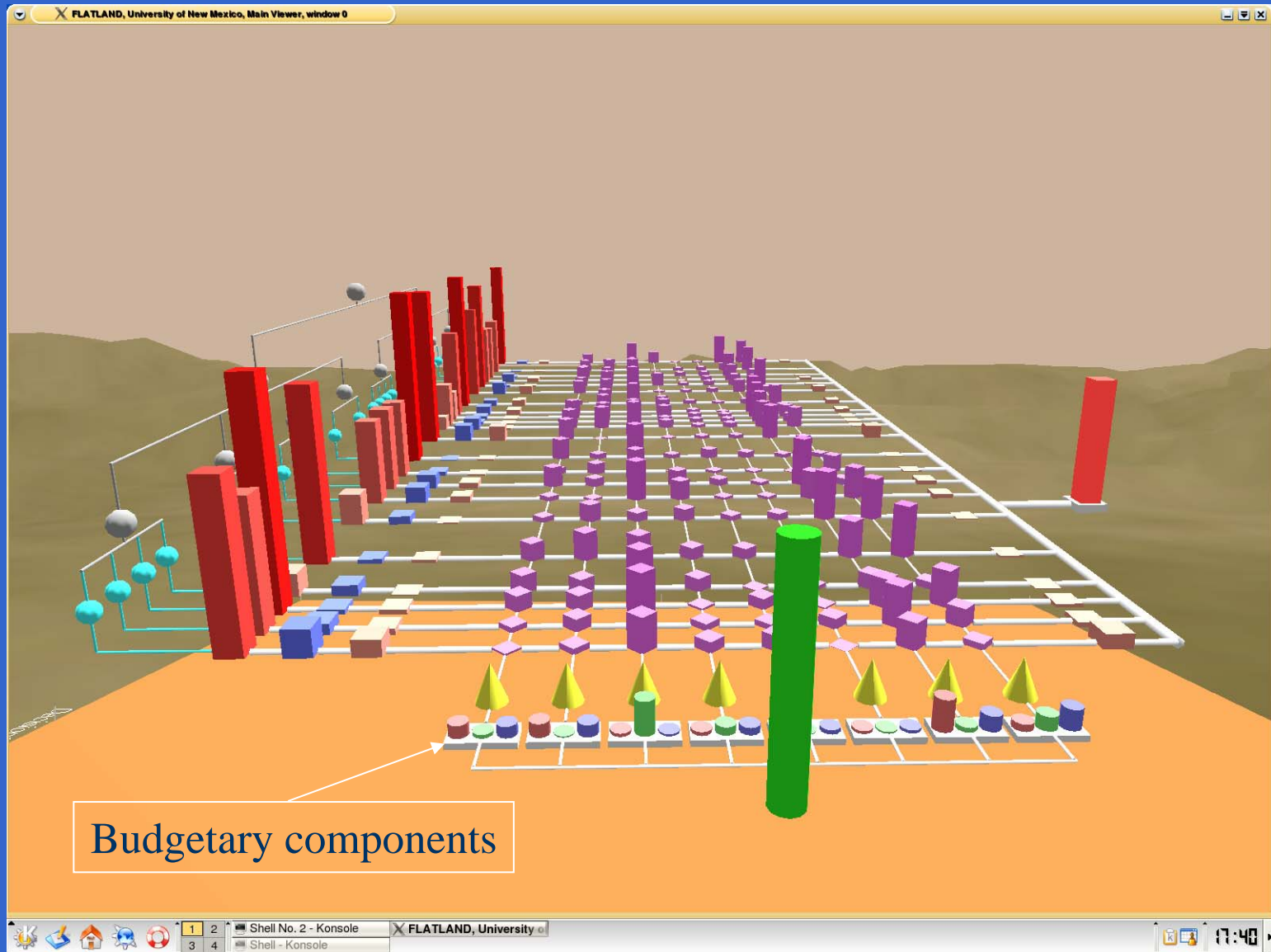
Visualization of Mockup System



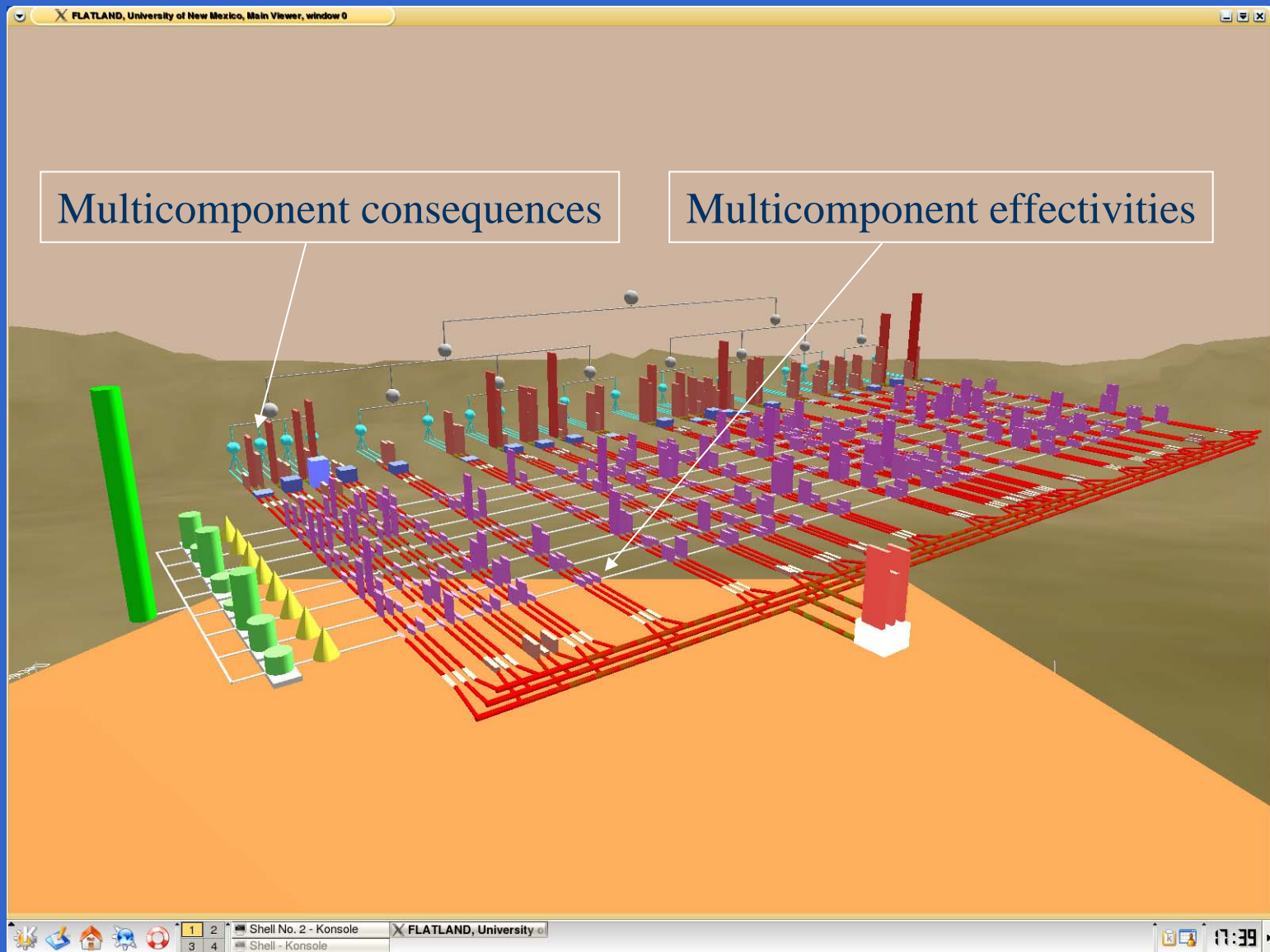
Visualization of Mockup System



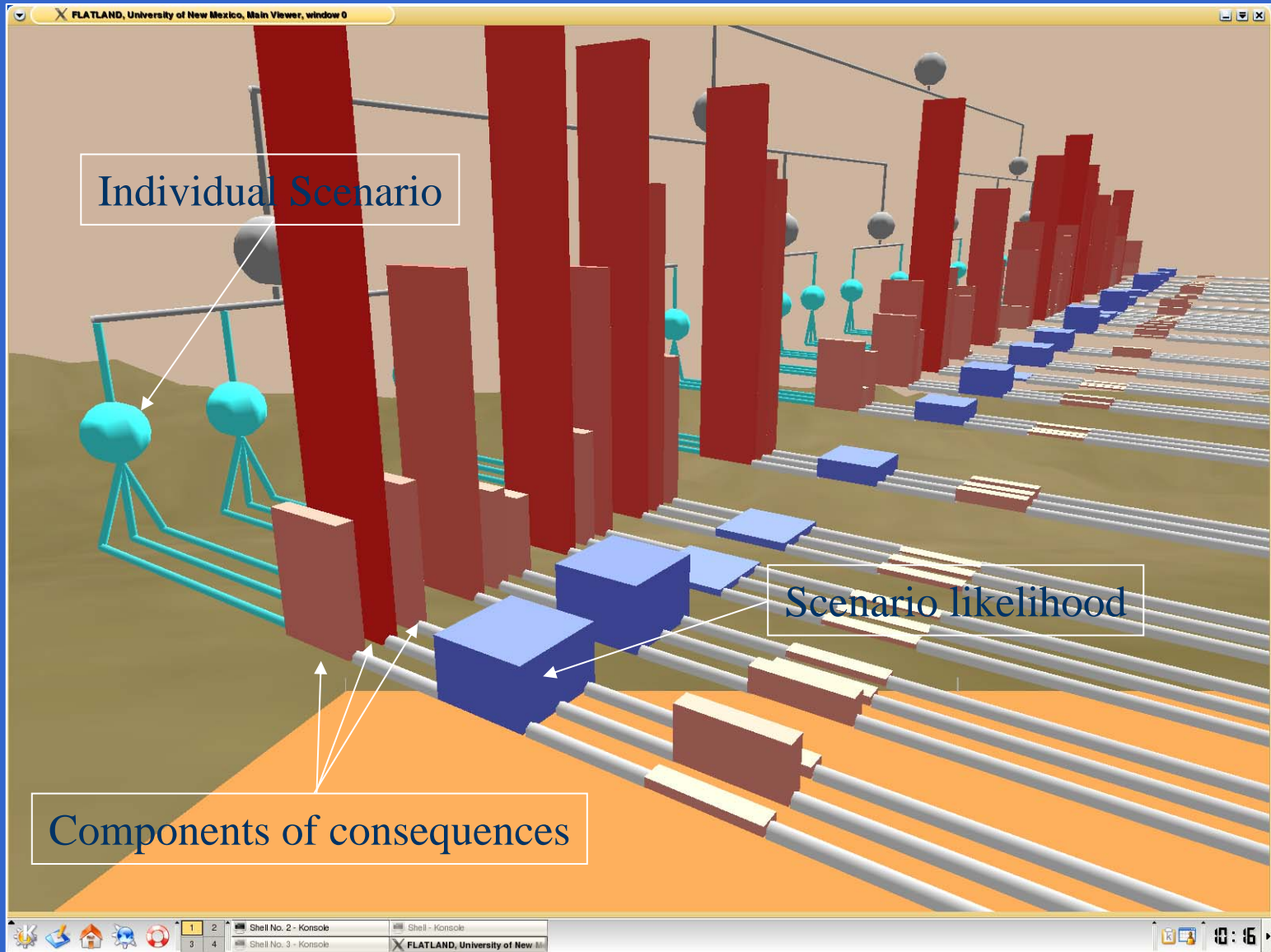
Visualization of Mockup System



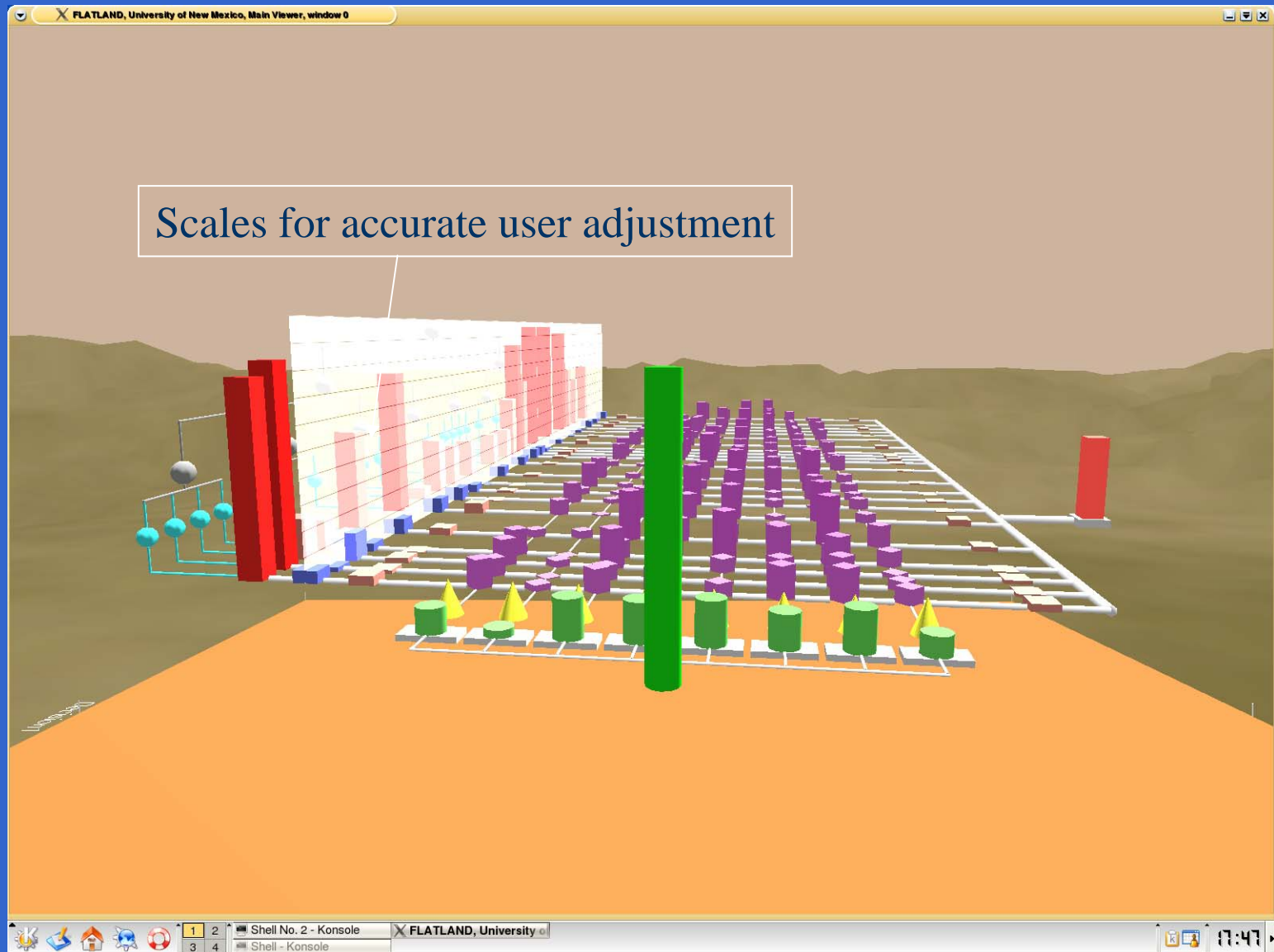
Visualization of Mockup System



Visualization of Mockup System



Visualization of Mockup System



Future advancements

- Integration with models and optimization,
- Integration with scenarios,
- Scalable scenario & remediation representations,
- Effectivity model representations,
- Quantitative measures of performance.

Invitation to demonstration
at the
University of New Mexico
Visualization Laboratory
Center for High Performance Computing

- Thursday -

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