

# **Road Safety Data Analysis**

## **Conflicting Objectives - Modeling Implications**

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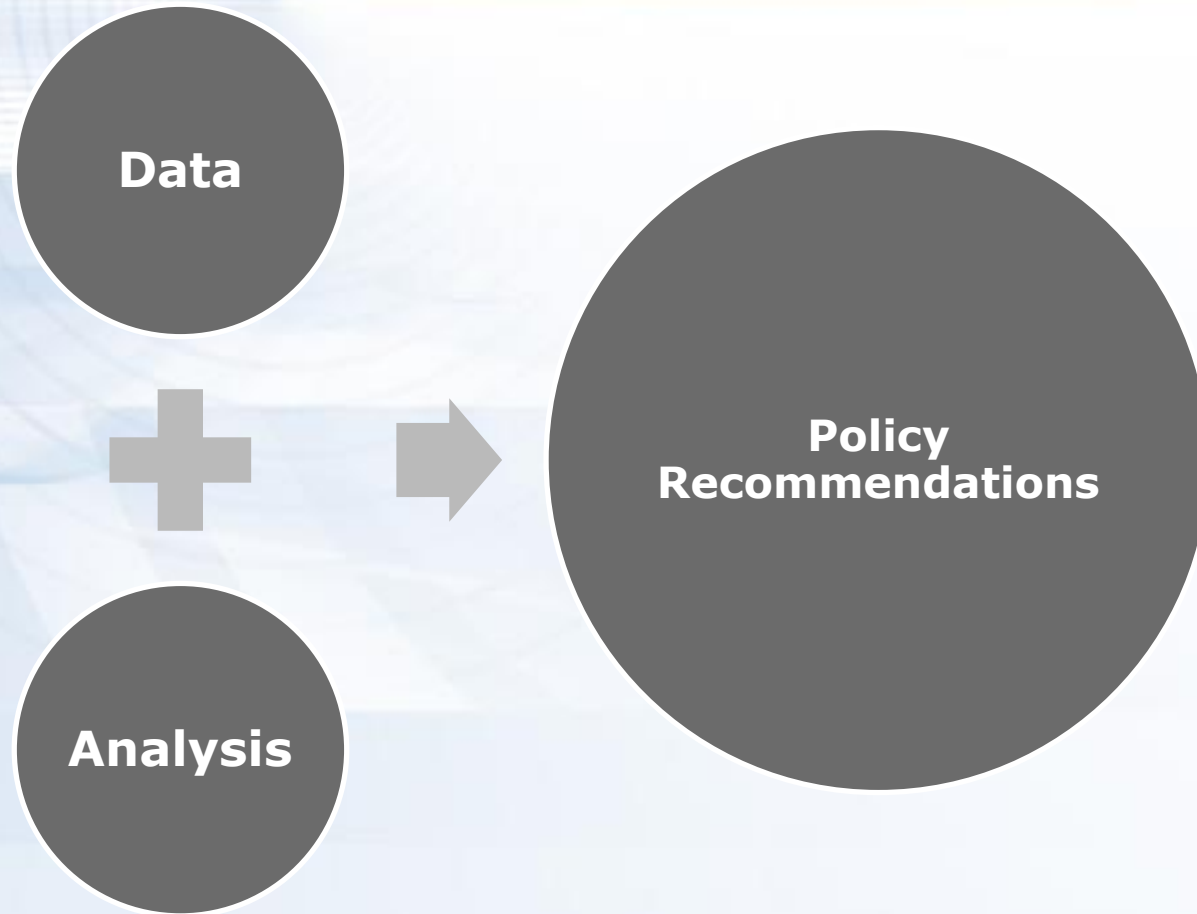
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# Outline

- 1. How are decisions made?**
- 2. New data in Safety**
- 3. Emerging modeling challenges**

# Data Analysis in Safety



*Does this relationship (always) work?*

# But...

- ❑ **Analyses may lead to “erroneous” policies**

1. Errors in data
2. Model misspecification
3. Complexity and hidden relationships

# (1) Data Errors

## ❑ Manual data collection

- ✓ police reports problematic (severity assessment, location)

## ❑ Incomplete datasets

- ✓ Lack of exposure data – insufficient information

## ❑ Spatial and temporal coverage

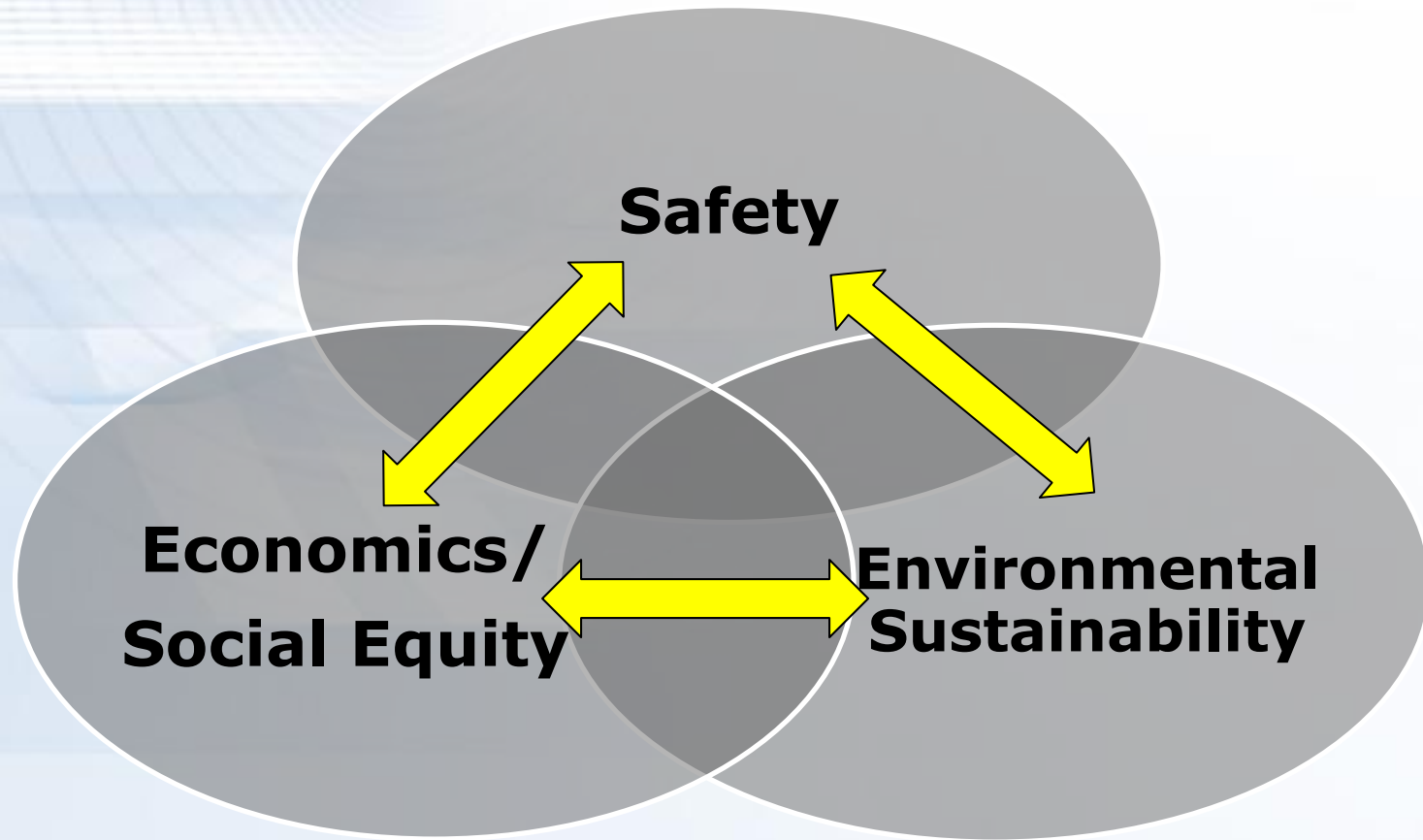
- ✓ Pre-, during, and post- crash data

## **(2) Model Misspecification**

### **□ Unreliable statistical analyses**

- ✓ Model selection
- ✓ Parsimonious vs fully specified models
  - **Spatial- and temporal correlations**
  - **Unobserved heterogeneity**
  - **Selectivity, bias/Endogeneity**
  - **Under-reported phenomena (severe crashes)**
  - **Risk compensation**

## (3) Complex Relationships



**Conflicting objectives?**

# Conflicting Objectives

## □ Environmental sustainability → highway safety

- ✓ Improve fuel efficiency and reduce GHG → smaller vehicles
- ✓ Climate change → crash frequencies and severities

## □ Highway safety → environmental sustainability

- ✓ Life cycle costs of road safety countermeasures (pavement rehabilitation, friction enhancements, ...)



# Conflicting Objectives

## □ Highway safety → social equity

- ✓ Increased costs associated with advanced safety features
- ✓ Vehicle safety features impose externalities on those without

## □ Economics/social equity → Highway safety

- ✓ Demand for freight movement → increases truck traffic & speed
- ✓ Increased mobility → traffic speeds & volume

# Understanding Complexities

- ❑ Many traditional tools are applicable (*efficient?*)
- ❑ Understanding of the process has been data-limited
- ❑ Driving Forces for Change
  1. New Technologies and Data Sources
  2. Novel Modeling Paradigms

**... where Change lies, Challenges Emerge!**

# Emerging Data Sources

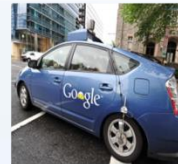
## □ Safety- Related

- ✓ More detailed crash and injury data
- ✓ Naturalistic driving data
- ✓ Vehicle Event Data Recorders (EDRs)

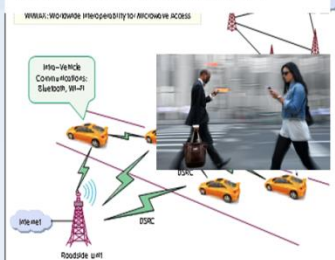
Mobile Internet



Autonomous vehicles



Wireless communications



Energy Storage Technologies



**BIG Data**

# Emerging Data Sources: Big Data

**Policy decisions with big data is ... different**

**❑ Much more operational data**

- ✓ Volumes, speeds, travel time etc.

**❑ Much more unstructured data**

- ✓ social media data
- ✓ Web transactions
- ✓ Smart objects



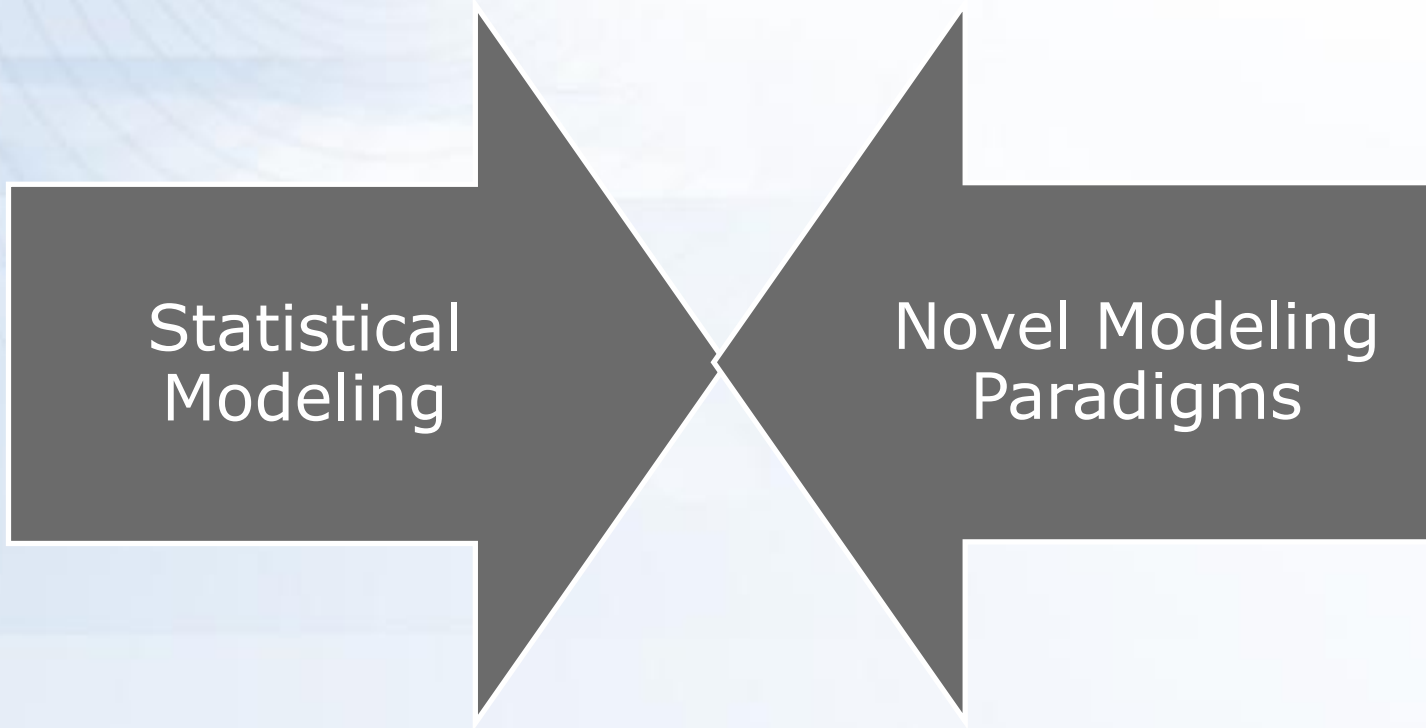
# Emerging Data Sources: Challenges

- 1. Database size and high dimensionality**
- 2. Overfitting and assessing statistical fit**
- 3. Rapidly changing and imperfect data**
- 4. Complexities and interactions**

**Can traditional statistical modeling efficiently transform massive data into knowledge?**

# Big Data Analysis

**Debatable...**



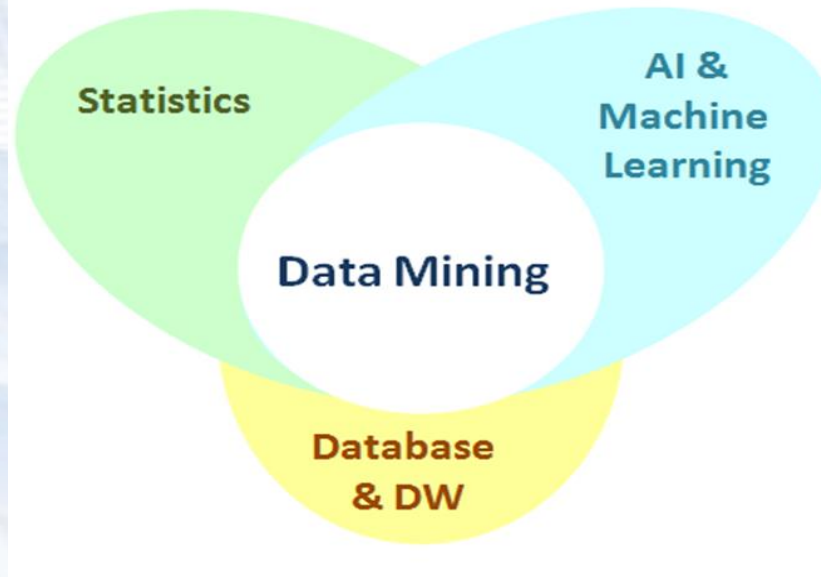
# Analysis: Statistical Modeling

- ❑ **Classical models**

- ❑ **But we also have extensions...**

- ✓ Random Parameter models
- ✓ Finite mixture latent class
- ✓ Markov switching
- ✓ Multivariate models
- ✓ Psychometric (ICLV)
- ✓ ...

# Analysis: Data Mining



- ❑ **Classical techniques from statistics, machine learning and CI**
- ❑ **Faster analysis of growing data with errors and imperfections**
- ❑ **Enables (easier) data fusion**
- ❑ **Robust to uncertainty**



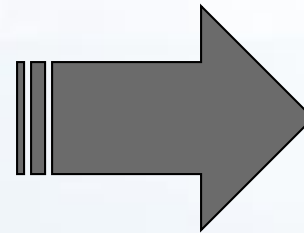
# Data Mining in Transportation

Three promising and rapidly growing research fields

1. ITS Applications

2. User Experience Mining

3. Behavioral Recognition



**Significant  
implications to  
proactive safety  
policies**

**The road ahead contains well  
developed statistical models  
and intelligent data mining**

....but with caution

# Key Drivers

- ❑ **Demystifying the gap between modeling and policy making**
  - ✓ Caveats
  
- ❑ **Unraveling the complex relationships**
  - ✓ Emerging technologies and novel methodologies
  - ✓ Means for success
    - **Synergies of statistics and Data Mining**
    - **Testbeds and Test data**

# Key Drivers

## Efficient policy recommendation will stem from understanding complexities

### ❑ Disregard complexities

- ✓ We will be simply ignoring potentially important information moving blindly forward

### ❑ Do something about them but poorly

- ✓ Will lead to erroneous and ineffective policies that will adversely impact safety, social equity and environmental sustainability

# A New Era in Safety Analysis

- ❑ **Safety will always be the cornerstone of modern societies**
- ❑ **Emerging technologies**
- ❑ **New and exciting data**
- ❑ **We also must consider**
  - ✓ Modeling Efficiency
  - ✓ Informed Decision Making
  - ✓ Proactive Management



# Some Thoughts

- ❑ **We need to expand our educational boundaries**
- ❑ **New Learning Paradigms**
  - ✓ Formulating new problems - revisiting old
- ❑ **New tools for dealing with old and new problems**
- ❑ **Beware of the hype...**

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