

# VIDEO QUALITY IMPACTS VIEWER BEHAVIOR: INFERRING CAUSALITY USING QUASI- EXPERIMENTAL DESIGNS

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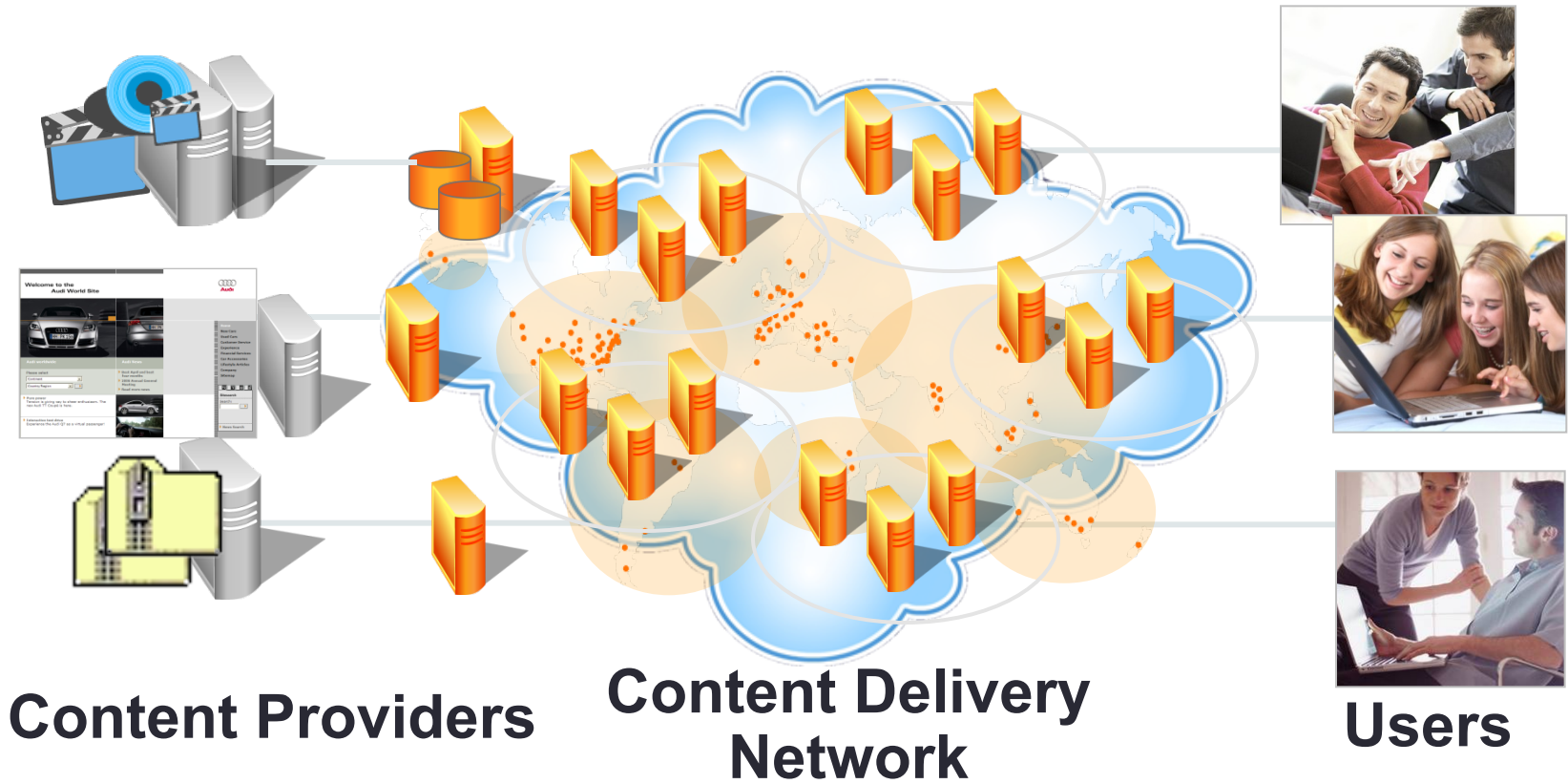
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Joint work with S. Shunmuga Krishnan (Akamai).

# The Video Delivery Ecosystem

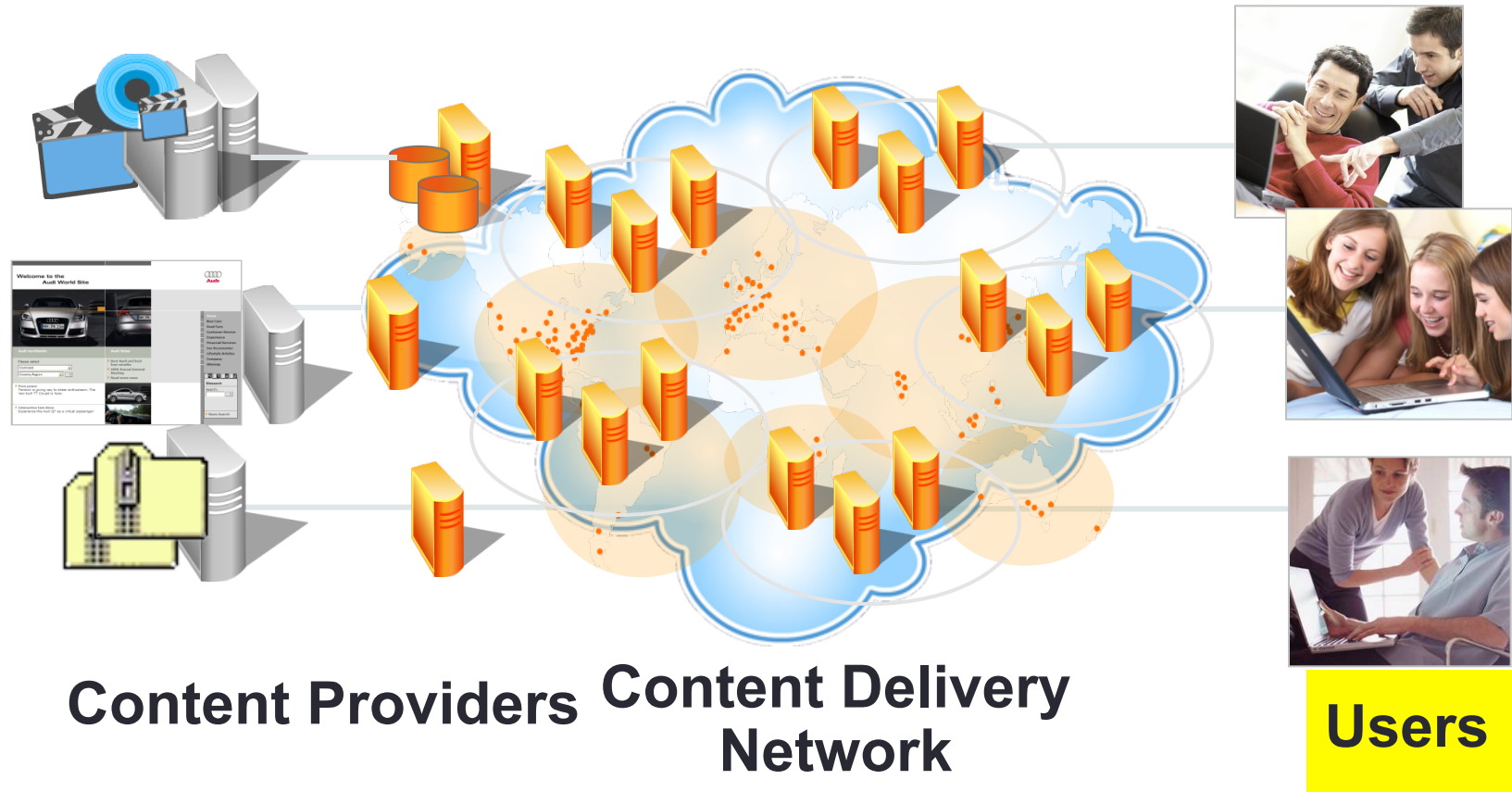


# The Video Delivery Ecosystem



- Media Providers: News, Movies, Entertainment, Sports, Television, ...

# The Video Delivery Ecosystem



- Different devices (desktop, mobile,...)
- Different geographies
- Different connectivity (cellular, DSL, cable, fiber, ...)

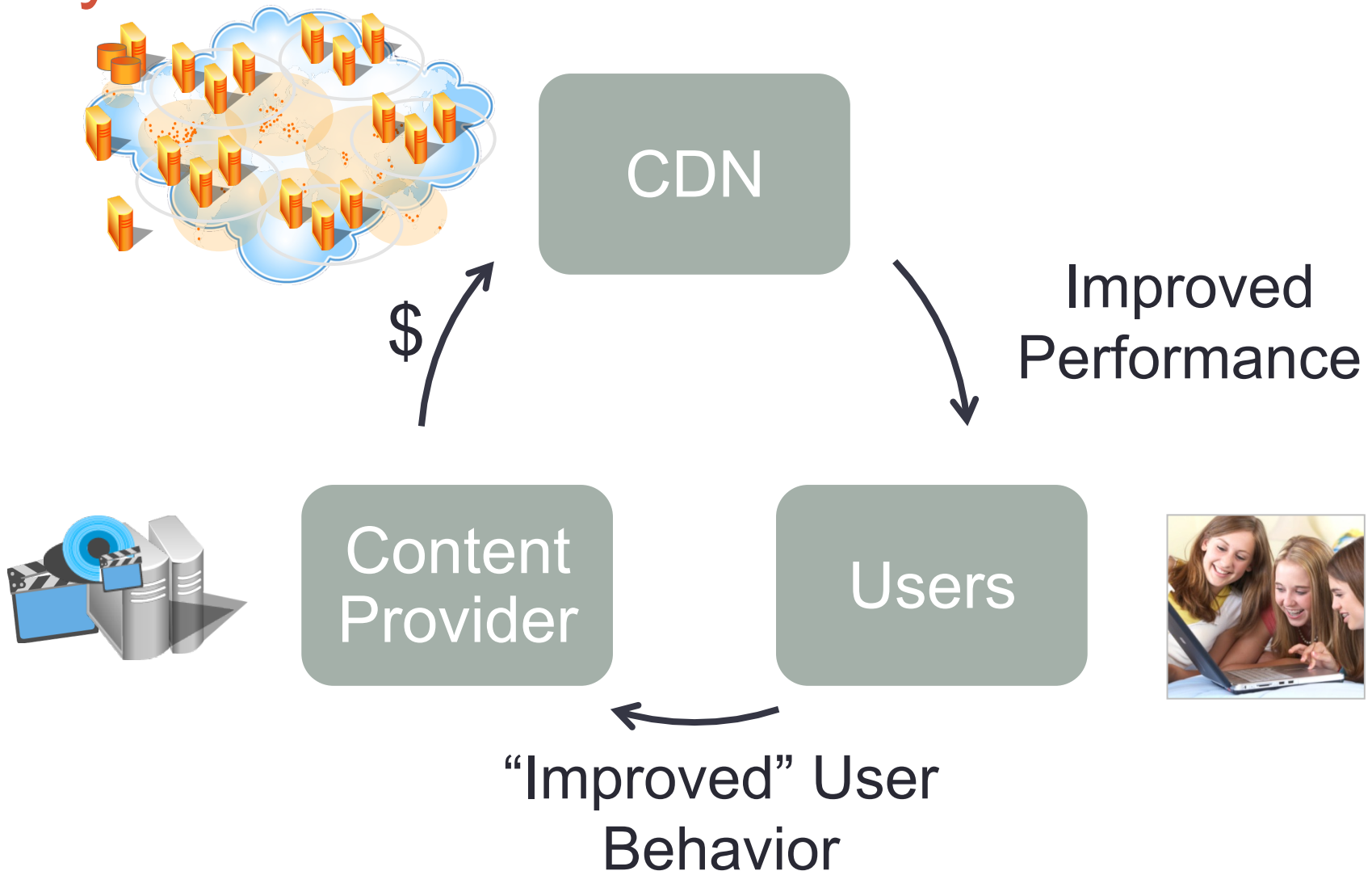
# The Video Delivery Ecosystem



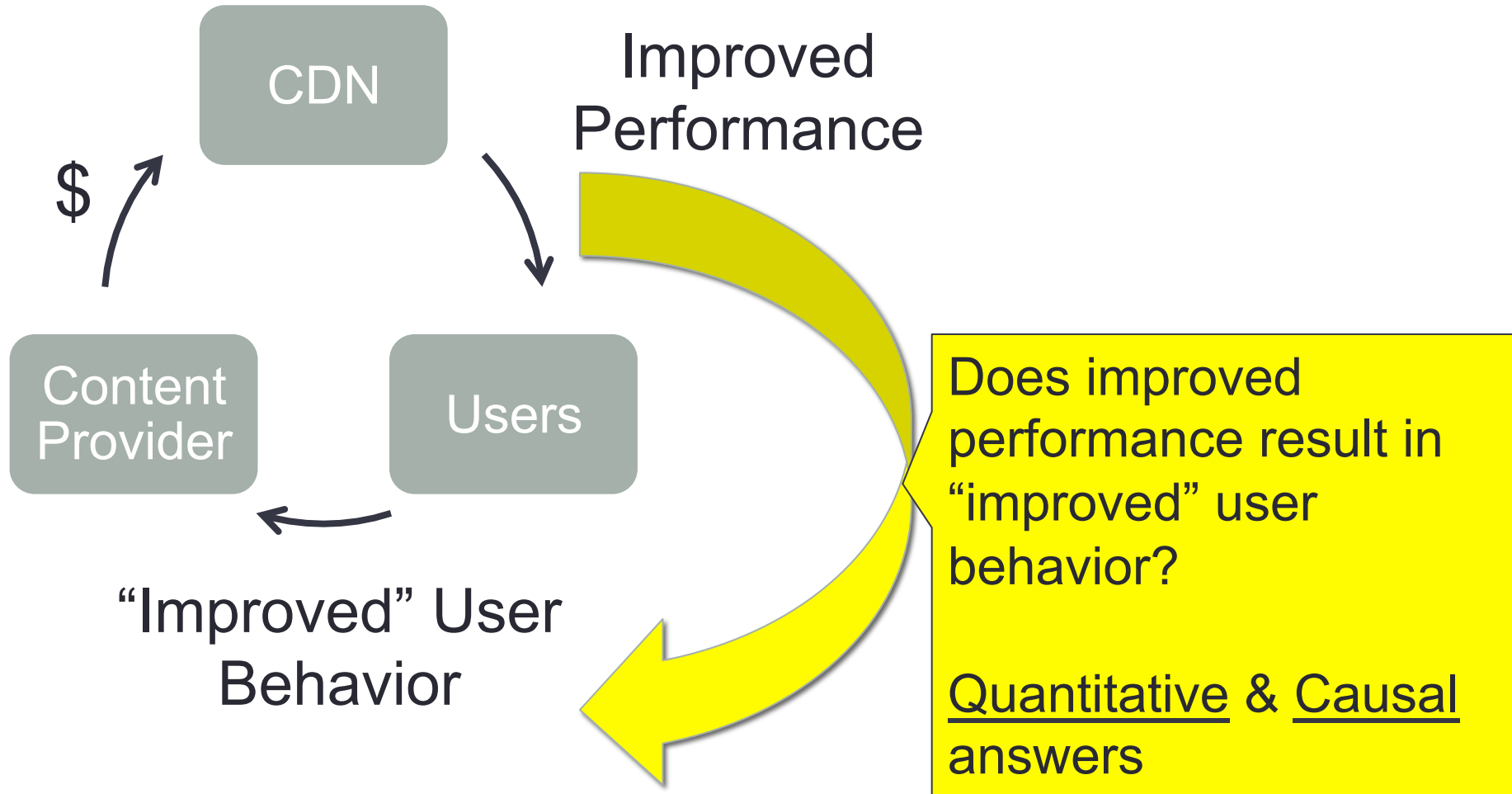
## Example: Akamai Network

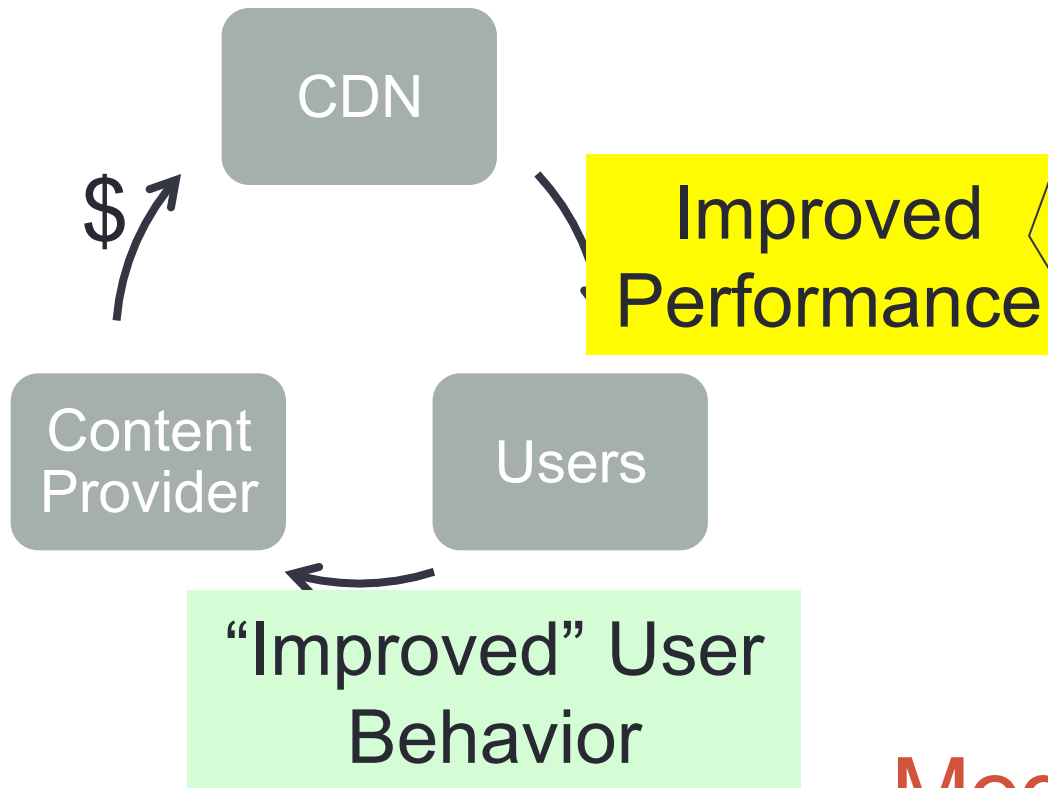
- 100,000+ servers in 1000+ clusters in 1000+ networks in 70+ countries serving trillions of requests a day.

# Video Delivery Economics: The Virtuous Cycle



# The Most Important and Least Understood Link





1. Availability: Viewers download video without failure.
2. Startup Delay: Video starts without much delay.
3. Rebuffers: Video plays without freezes.

## Media & Entertainment (Ad-supported or Subscription)

1. Abandonment: Reduce viewers who abandon without viewing the video.
2. Engagement: Viewers watch videos longer.
3. Repeat Viewership: Viewers keep coming back to site to watch more videos.

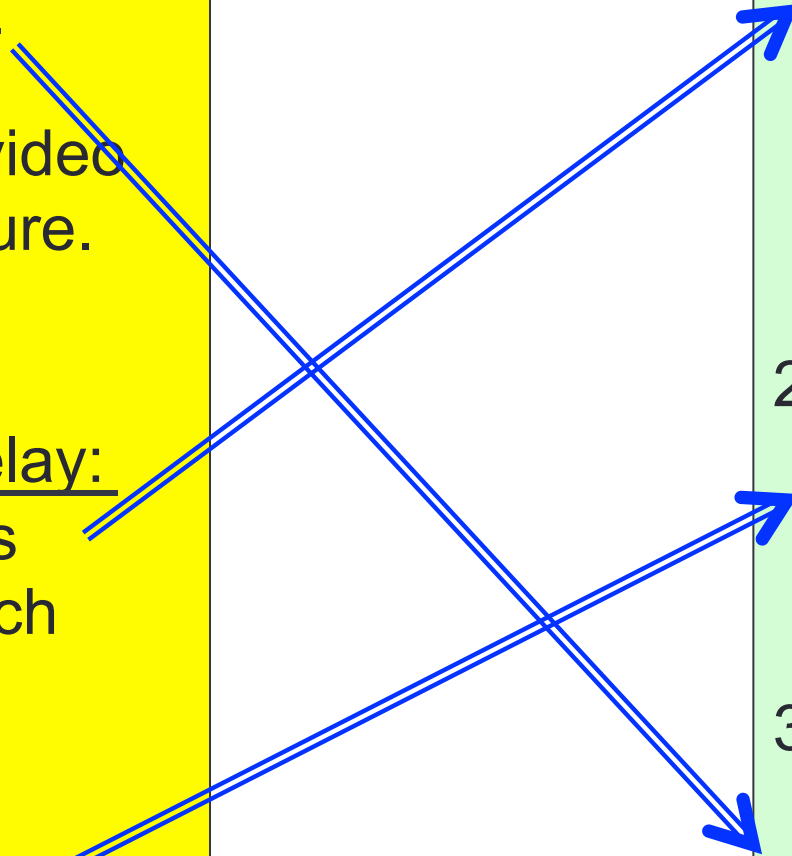


## VIDEO PERFORMANCE

1. Availability:  
Viewers download video without failure.
2. Startup Delay:  
Video starts without much delay.
3. Rebuffers:  
Video plays without freezes.

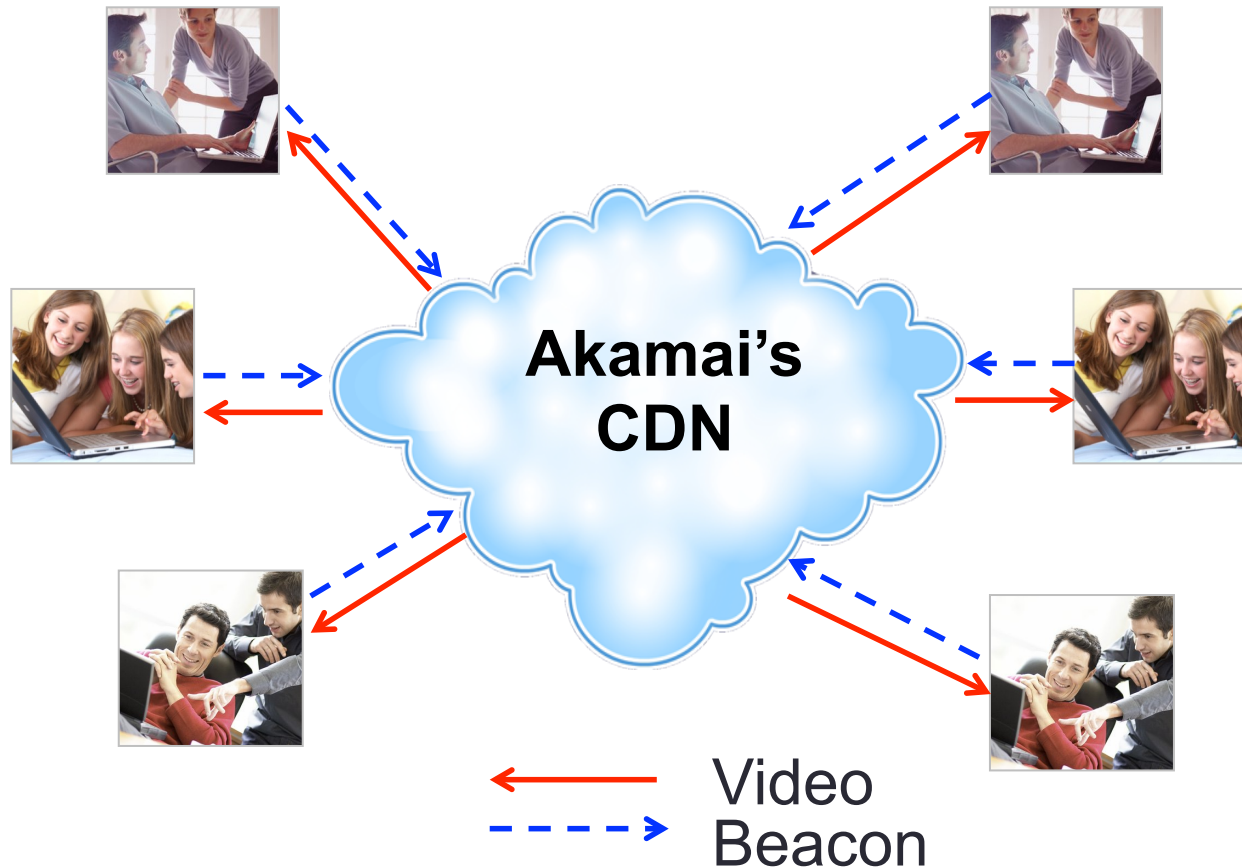
## VIEWER BEHAVIOR

1. Abandonment:  
Reduce viewers who abandon without viewing the video.
2. Engagement:  
Viewers watch videos longer.
3. Repeat Viewership:  
Viewers keep coming back to site to watch more videos.



# The Data

# Akamai's Client-side Player Plugin



Globally-deployed Akamai plugin that runs inside the media player and reports viewer actions and performance metrics via ``beacons'' from millions of actual end-users around the world.

# Our Data Set

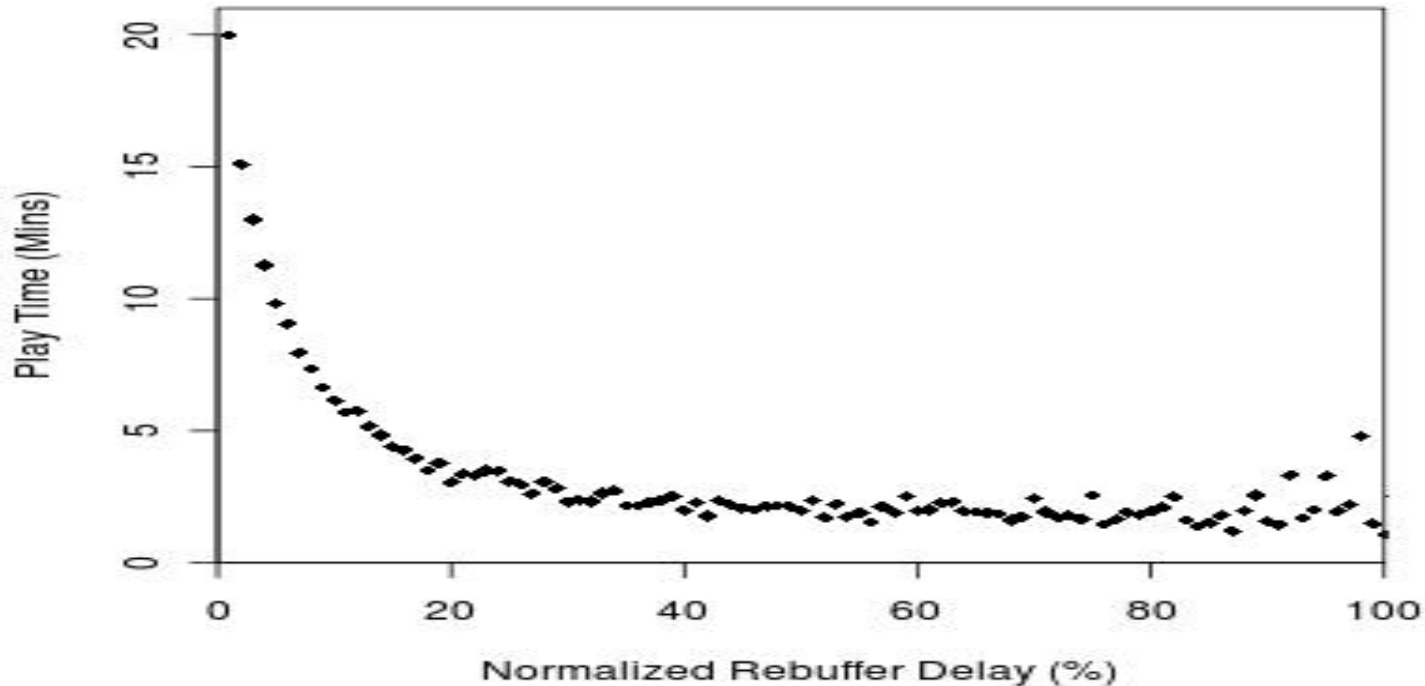
One of the most extensive data sets *ever* analyzed for this purpose.

Analyzed data from the widely-deployed Akamai's client-side plug-in.

- 6.7 million unique viewers
- 23 million views
- 216 million minutes of video played
- 102 thousand unique videos
- Viewers in three continents (NA, Europe, and Asia)

# The Techniques

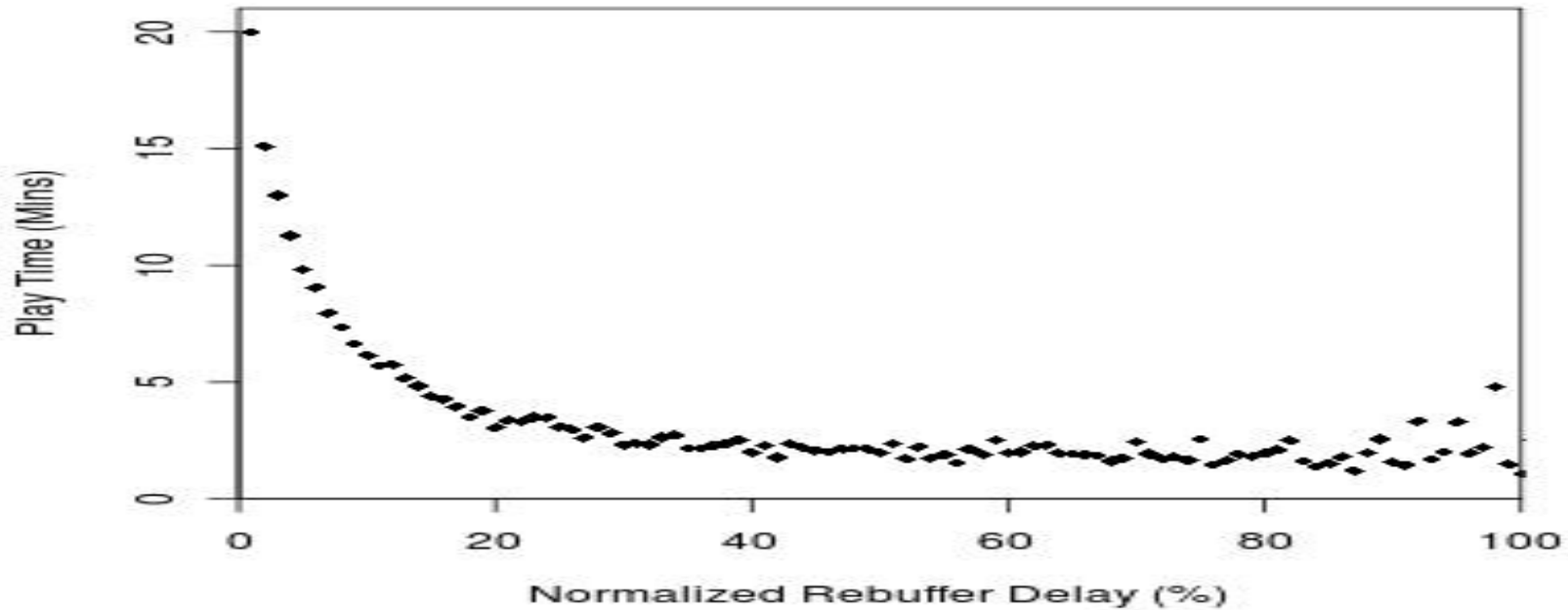
# Correlation



Hypothesis: Video rebuffering causes viewers to watch less.

Strong negative rank correlation. Kendall correlation = -0.421.

# Correlation



Threat to causality: Users who are better off can afford better network **connectivity**, resulting in less rebuffering. They can also afford access to more interesting **content**.

# Correlation $\neq$ Causality

Correlation: A and B “move together”.

versus

Causality: A causes B to occur.

Threats to Causality: **Confounding variables** that could account for both A and B.

Typical confounding variables: **Connectivity, Content, Geography**.



# Randomized Experiments

Idea: Equalize the impact of confounding variables using random assignment (Fisher 1937)

1. Randomly assign subjects to receive “treatment” A.
2. Compare outcomes of treatment versus the “untreated” control group.

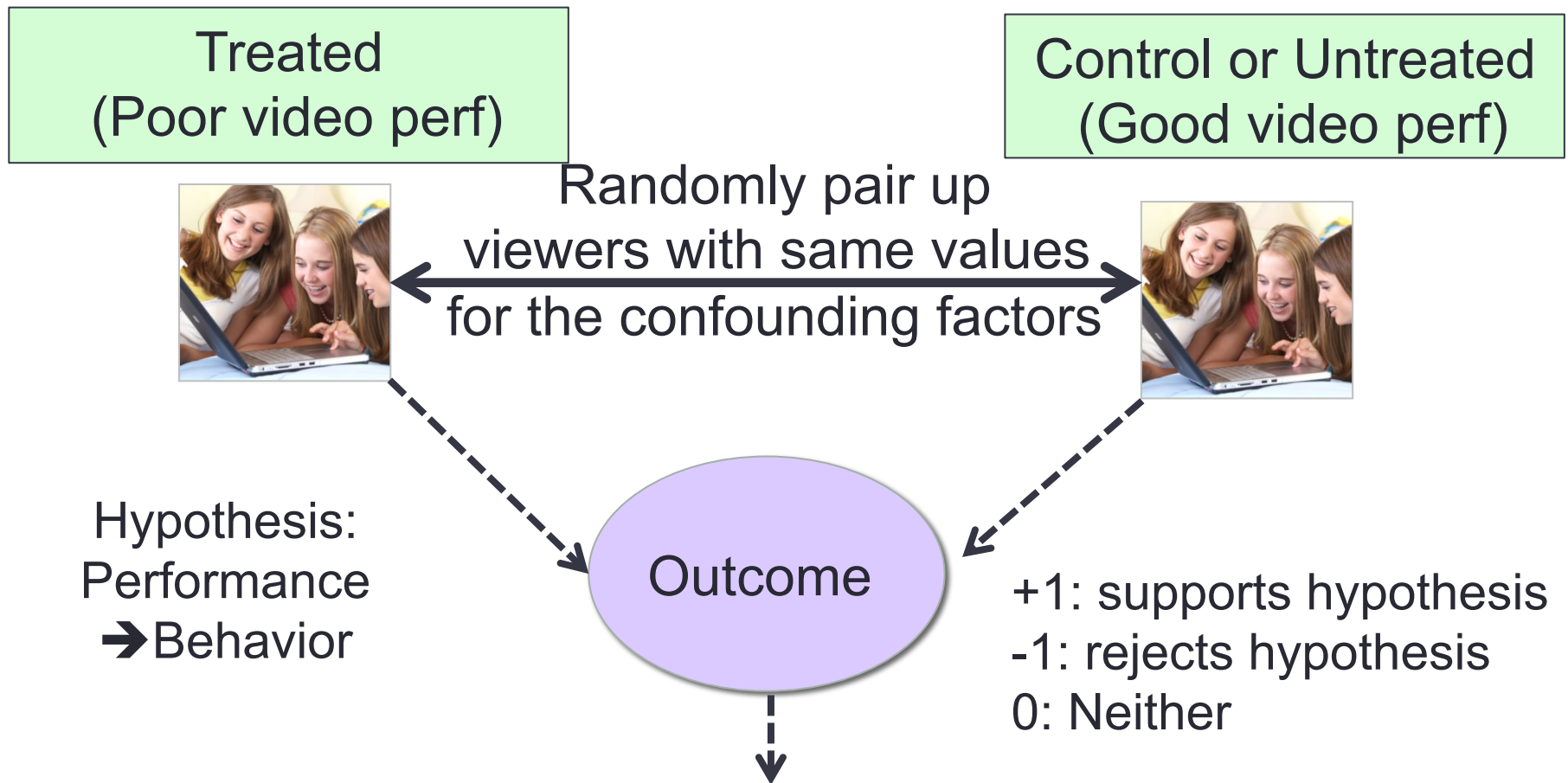
Treatment = Degradation in Video Performance

Hard to do:

Operationally  
Cost Effectively  
Legally  
Ethically

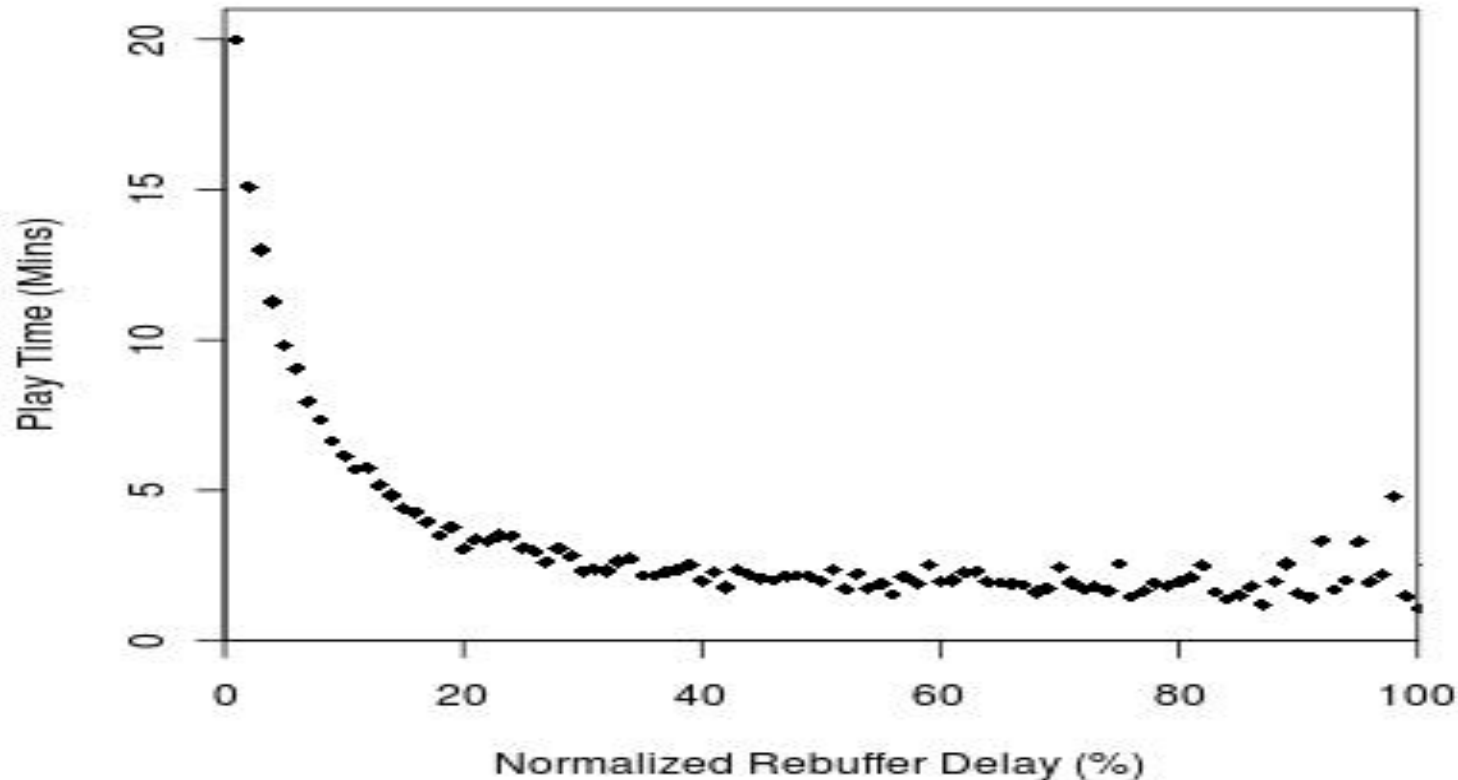
# Our Approach: Quasi-Experiments

Idea: Equalize confounding factors by *experimental design*.  
Example, Matched design (Levy et al 1985 nutrition study)



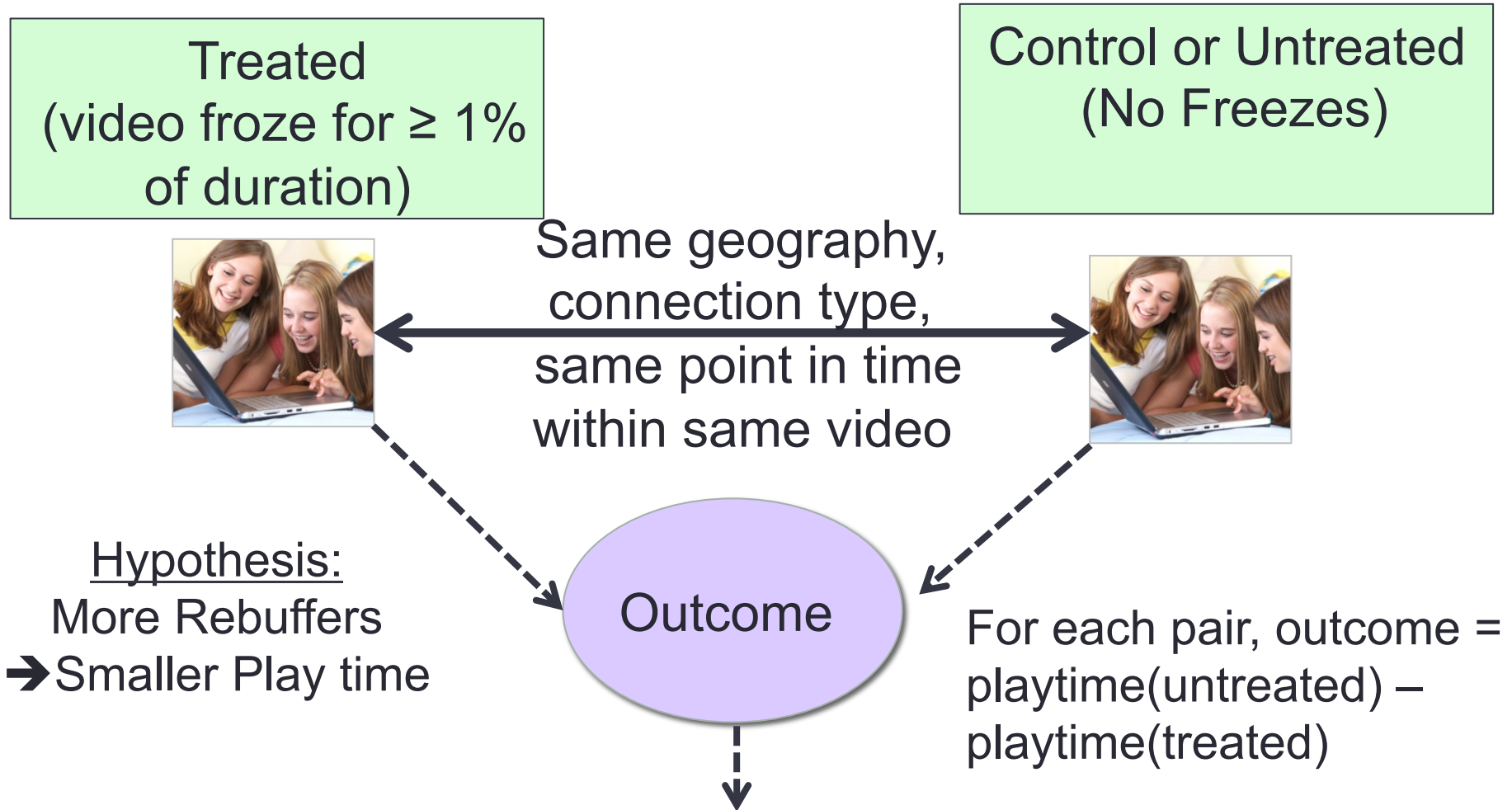
# Viewer Engagement

# Does rebuffering reduce the average time a viewer plays a video?



Strong negative correlation (-0.421): increased normalized rebuffer delay correlates with decreased play time.

# Quasi-Experiment for Viewer Engagement



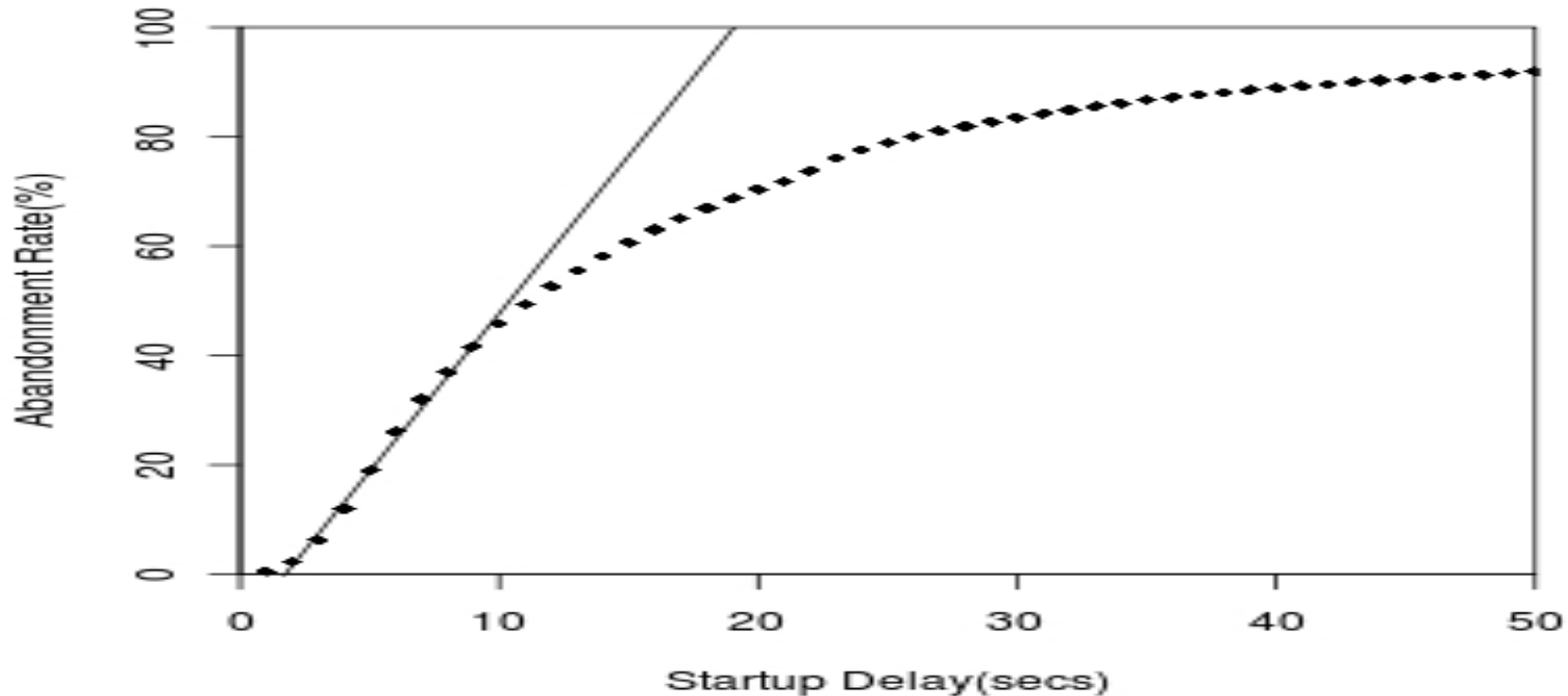
## Results of Quasi-Experiment

Normalized Rebuffer Delay ( $\gamma\%$ )	Net Outcome
1	5.0%
2	5.5%
3	5.7%
4	6.7%
5	6.3%
6	7.4%
7	7.5%

A viewer experiencing rebuffering for 1% of the video duration watched 5% less of the video compared to an identical viewer who experienced no rebuffering.

# Viewer Abandonment

# How long will viewers wait for a video to startup?



$\text{AbandonRate}(x) = \% \text{ of views abandoned if startup delay is } z$   
 $= 100 \times (\text{Impatient}(x) / (\text{Impatient}(x) + \text{Patient}(x)))$ .

- Viewers start to abandon if startup delay exceeds 2 seconds.
- Beyond 2 seconds, a 1-second increase in delay results in roughly a 5.8% increase in abandonment rate.



## What is more frustrating?

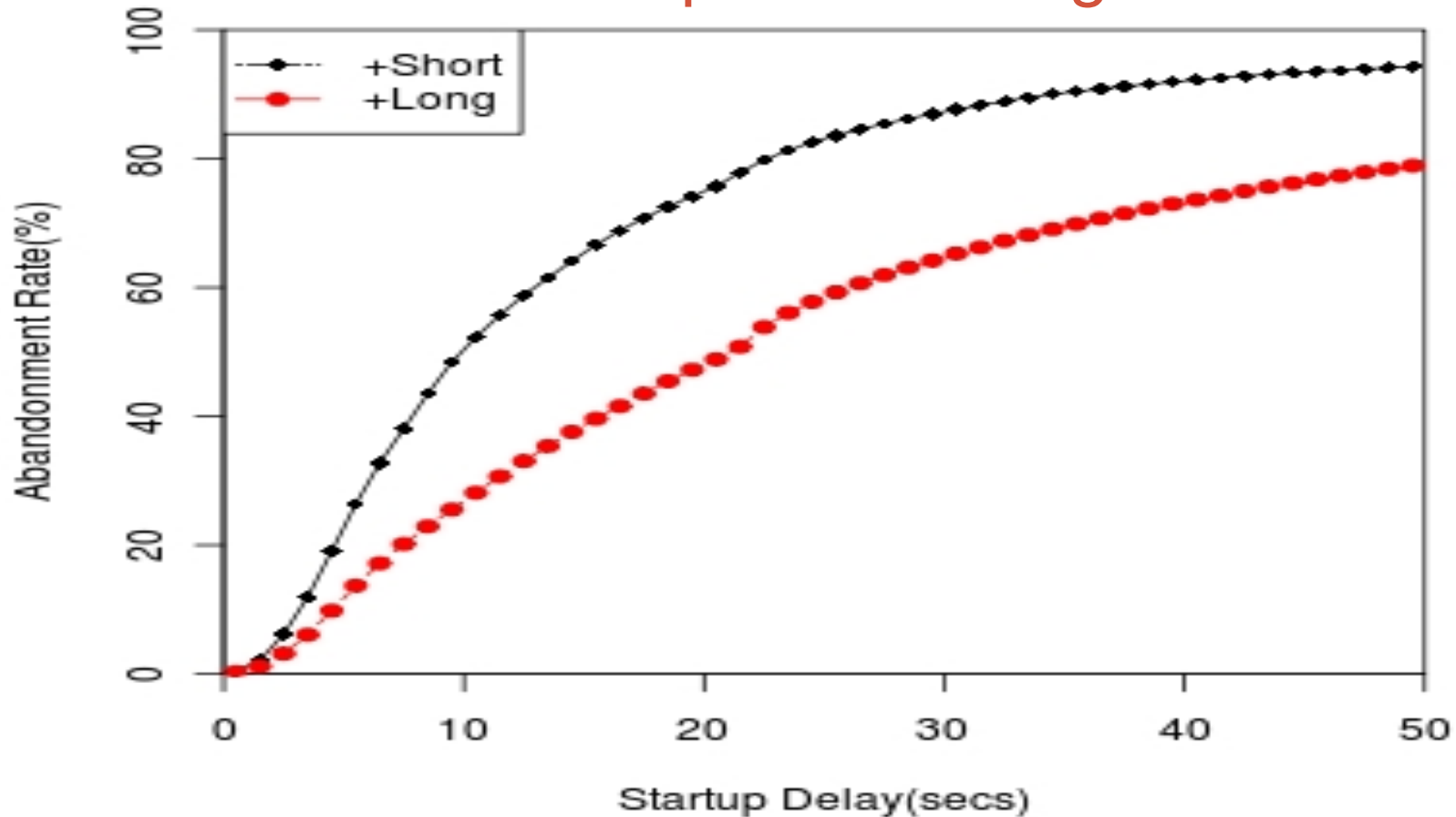
Waiting 30 minutes for a long plane ride?



Waiting 30 minutes for a short cab ride?



## Viewers are less tolerant of startup delay for short videos in comparison to longer videos



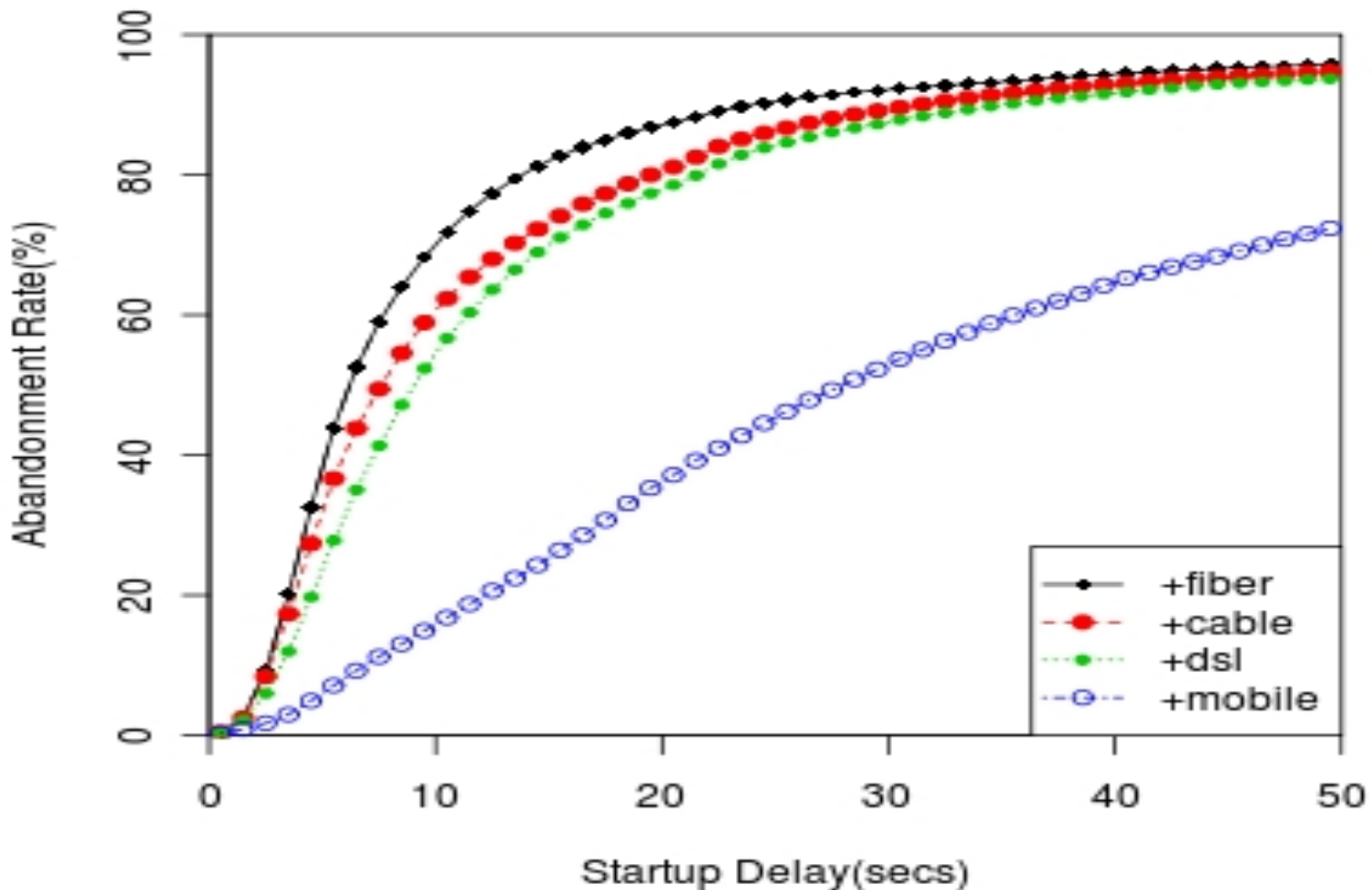
Short: < 30 mins (e.g, news clip). Median Duration: 1.8 mins  
Long:  $\geq$  30 mins (e.g, movie). Median Duration: 43.2 mins

## Anyone for the Lightning Express?



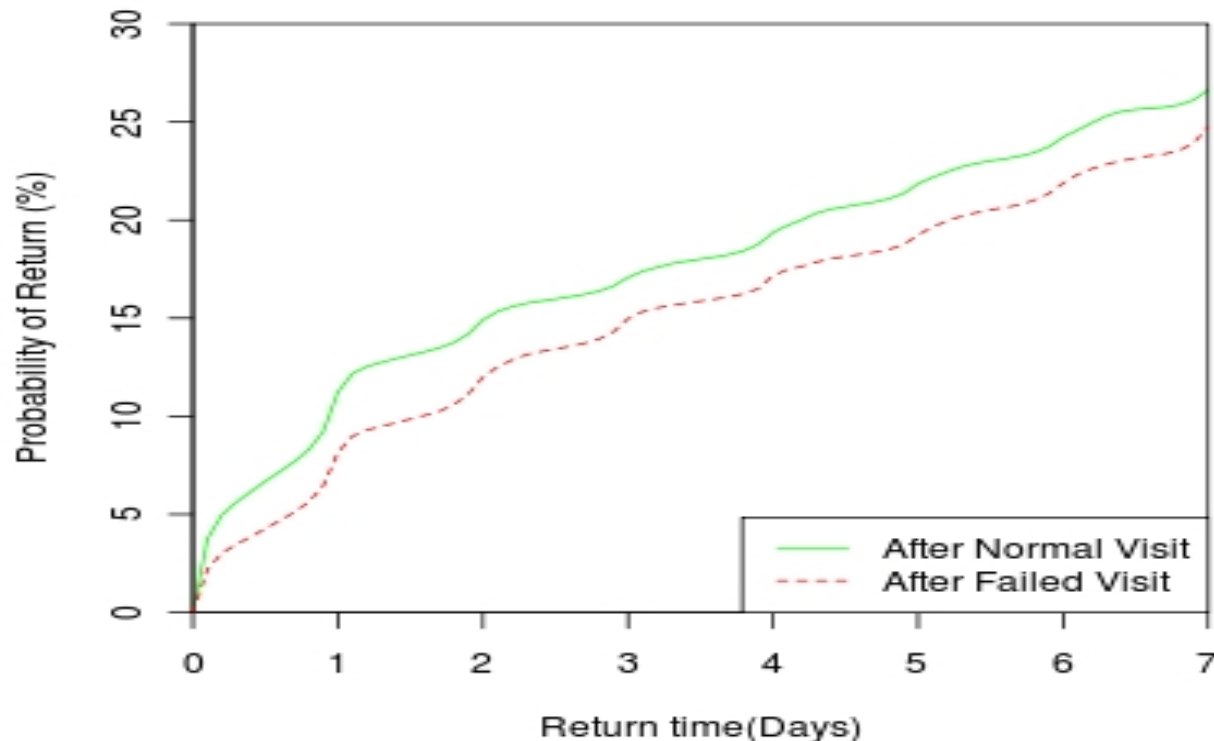
“Express train crosses the nation in 83 hours.”  
New York Times, June 4<sup>th</sup> 1876.

Viewers with better connectivity have less patience for startup delay and abandon sooner.



# Repeat Viewership

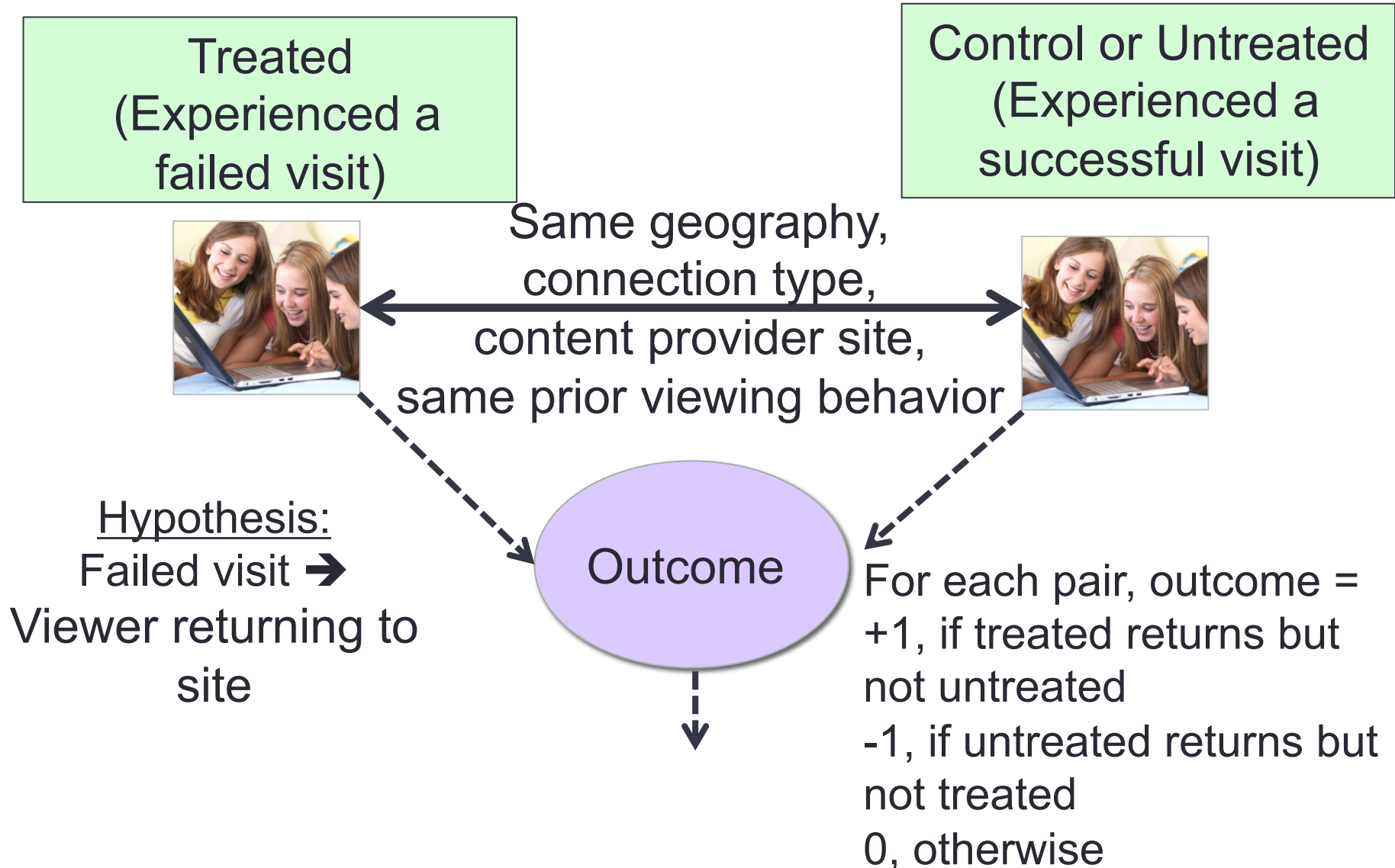
# Do failures reduce the likelihood that a user will return to the same content provider's site?



Failed Visit= The viewer fails to play a video and leaves the site ending the visit.

The probability of a viewer returning to a content provider's site within a specified time is distinctly smaller after a failed visit than after a normal visit.

# Quasi-Experiment for Repeat Viewership



# Results from Quasi-Experiment

A viewer experiencing a failed visit is 2.32% less likely to return to the same content provider's site within a week than a similar viewer that had a successful visit.



# Our Contributions

First large-scale quantitative study of the causal link between video performance and viewer behavior.

- Prior work: correlational study of viewer engagement (Dobrian et al 2011).

Deeper and better understanding of

- how to architect delivery networks (for architects)
- user behavior and video monetization (for content providers)

New Quasi-Experimental Design (QED) techniques for causal inference in network measurement.

- Prior work: QED in social and medical sciences but not in our domain.

Questions?