

# Analysis of Peer-Assisted Video-on-Demand Systems with Scalable Video Streams

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**(Joint work with Kianoosh Mokhtarian)**

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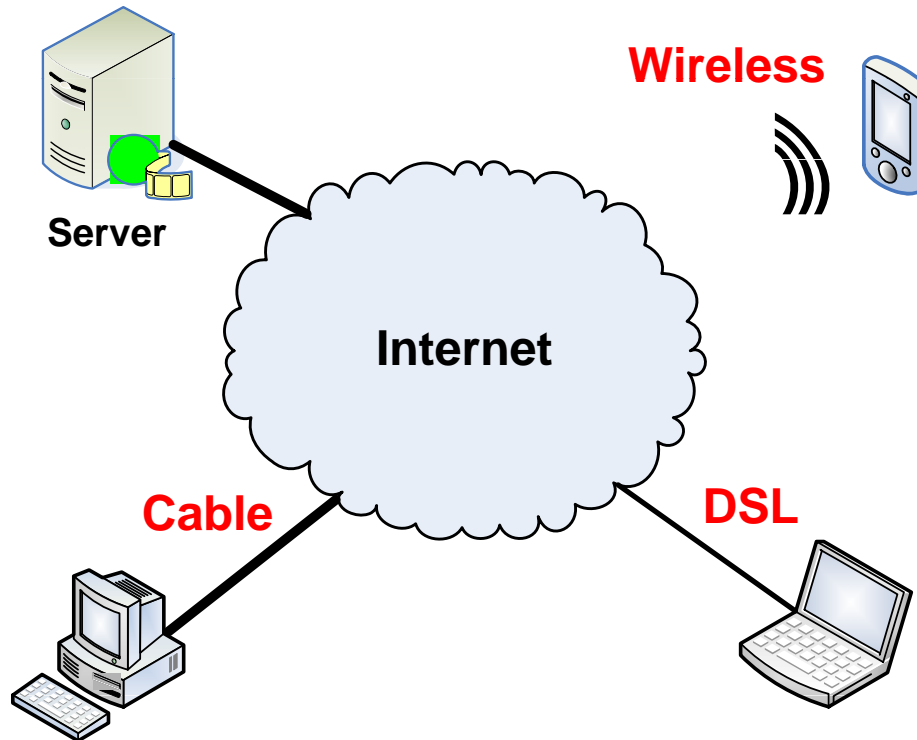
# Motivations

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- **Wide deployment of P2P streaming systems**
  - PPLive, UUSee, SopCast, CoolStreaming, ....
- **Video streams: 300 kbps to 1.5+ Mbps**
- **Typically more than upload bandwidth of most peers → seed servers needed to help → peer-assisted streaming systems**

# Motivations

- **Receivers are quite heterogeneous**
  - network, processing power, screen resolution, ...



# Supporting Receiver Heterogeneity

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- **Encode low bitrate videos**
  - Supports more receivers
  - Low video quality for everyone
- **Distribute multiple versions (simulcasting)**
  - Need to encode video several times
  - Switching among versions (wait for next I-frame)
  - Reduced connectivity (P2P): different stream versions → different P2P networks

# Supporting Receiver Heterogeneity

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- **Multiple Description Coding (MDC)**
  - Video quality ~ number of descriptions received
  - Considerable bitrate overhead
  - Computationally complex
- **Scalable Video Coding (SVC)**
  - Encode and distribute one video stream
  - Extract and decode various substreams
  - Lower overhead, simpler than MDC
  - **Our focus in this paper**

# Our Contributions

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- **Analytical model for peer-assisted streaming systems with SVC**
  - General model; details worked out for typical systems
  - Validated using simulations
- **Provides insights and answers questions such as**
  - What is the expected throughput or video quality?
  - How much seeding capacity is needed to increase quality to certain level?
  - How many peers can the system support in flash crowd scenarios?

# Previous Works

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- **Capacity growth of P2P system over time [Tu 05]**
  - Finding the time to turn off servers
  - Similar analysis based on fluid model in [Kumar 07]
  - Only nonscalable video streams
- **Streaming capacity/startup delay and piece selection strategies [Zhou 07, Parvez 08]**
  - Random matching of peers
- **Analysis based on constructing specific overlay topology [Small 06, Liu 08]**
  - Cannot support most of today's systems

# System Model

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- **Trackers**

- Receive periodic update reports from peers

- **Peers**

- Join by contacting one of the trackers
- Try to serve lower layers first
  - **As many layers as they can**
- Download in a streaming form, at the video bitrate

- **Seed servers**

- Serve requests when not enough capacity in peers
- Finite resources



# System Model

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- **Video file divided into  $S$  small segments**
  - Time unit = segment length =  $T_{seg}$  (e.g., 1 minute)
  - Peers in  $P_s$  (peers currently watching segment  $s$ ) can serve to peers in  $P_{1,\dots,s-1}$
- **Serving requests by seed servers**
  - Random
  - Contribution based [Mokhtarian 09]

# System Model

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## ■ Inputs (characteristics of P2P system)

- Joint distribution of upload and download bandwidths (random variables **D** and **U**)
  - $Pr(x_1 \leq \mathbf{D} \leq x_2, y_1 \leq \mathbf{U} \leq y_2)$
- The expected peer seeding duration  $T_{seed}$
- Capacity of seed servers **C**, number of video layers **L**, bitrate of each layer  $r_l(t)$  at each segment  $t$
- Distribution of peer arrivals (random variable **N**)
  - $Pr(\mathbf{N} = n)$ : probability of  $n$  arrivals in one time unit ( $T_{seg}$ )
- Peer failure/departure rate  $\gamma(t)$  at segment  $t$

# System Model

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- **Outputs**

- **System throughput**
- **Average video quality served**
- **Max number of supported peers (flash crowd)**

# General Analytical Model: Overview

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- **High-level procedure**

- **Step 1:** Estimate the capacity that peers in  $P_s$  can serve to others (i.e.,  $P_{1,\dots,s-1}$ ) as a function of the capacity served to  $P_s$
- **Step 2:** Employ this function for  $s = S, S-1, \dots, 1$  to analyze the distribution of the complete video

# General Analytical Model: Step 1

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- **Step 1: Distribution of one video segment**
  - $f(e, l, r_l, n)$ : expected value of output capacity for layer  $l$  of a segment when serving it with capacity  $e$  to  $n$  peers
    - $e \geq r_l \times n$ : trivial  $\rightarrow f(\cdot)$  in this case denoted by  $f_{all}(\cdot)$
    - $e < r_l \times n$ : only  $\lfloor e/r_l \rfloor$  of the  $n$  peers can be served
      - Selected randomly  $\rightarrow f_{rnd}(\cdot)$
      - Selected based on contribution  $\rightarrow f_{opt}(\cdot)$
  - Auxiliary rnd variable  $U_l$ : upload bw of peers for layer  $l$ 
    - Distr of  $U_l$  estimated using  $U$  and peers upload behavior

# General Analytical Model: Step 1

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## ■ Step 1: Distribution of one video segment

-  $f(e, l, r_l, n)$ : expected value of the output capacity

$$f_{all}(e, l, r_l, n) = n \times E[ \mathbf{U}_l \mid \mathbf{D} \geq R_l ]$$

$$f_{rnd}(e, l, r_l, n) = k \times E[ \mathbf{U}_l \mid \mathbf{D} \geq R_l ]$$

- $k$ : num peers that capacity  $e$  can serve with layer  $l$
- $R_l$ : bitrate of the first  $l$  layers
- $E[ \mathbf{U}_l \mid \mathbf{D} \geq R_l ]$ : expected upload bw (for layer  $l$ ) of a peer given that it can download layers 1 to  $l$

$$E[ \mathbf{U}_l \mid \mathbf{D} \geq R_l ] = \frac{\int_0^\infty Pr(\mathbf{U}_l = x, \mathbf{D} \geq R_l) x dx}{Pr(\mathbf{D} \geq R_l)}$$

# General Analytical Model: Step 1

## ■ Step 1: Distribution of one video segment

- $f_{opt}(e, l, r_l, n)$ : expected output capacity of the first  $k$  peers out of a set of  $n$
- Assume a sorted list of peers according to serving capacity for layer  $l$ . Let  $k+1$ -st peers have capacity  $x$ 
  - The expected capacity of the first  $k$  peers:

$$E[k \times \mathbf{U}_l \mid \mathbf{U}_l > x, \mathbf{D} \geq R_l]$$

$$\Rightarrow f_{opt}(e, l, r_l, n) = \int_0^\infty E[k \times \mathbf{U}_l \mid \mathbf{U}_l > x, \mathbf{D} \geq R_l] \times$$

$$\Pr(k \text{ peers have capacity } > x, \text{ one has } x, n - k \text{ have } < x) dx$$

# General Analytical Model: Step 2

## Step 2: distribution of video segments over time

### - Peers dynamics (at a given segment)

- Using  $f_{opt|rnd|all}(\cdot)$  obtained so far, the expected output capacity of a set of peers at a segment is:

$$\sum_{n=0}^k \Pr(n \text{ peers demand } l) \times f_{all}(e, l, r_l, n) \times \Pr(\text{a peer stays}) +$$
$$\sum_{n=k+1}^{\infty} \Pr(n \text{ peers demand } l) \times f_{opt|rnd}(e, l, r_l, n) \times \Pr(\text{a peer stays})$$

- Pr(there are  $n$  peers demanding layer  $l$  at a given seg):  
calc'ed using distr of **N** & **D** (arrivals and bandwidths)
- Pr(a peer at a given seg stays until next seg):  
calculated using peer failure rate



# General Analytical Model: Step 2

## ■ Step 2: distribution of video segments over time

- We know how to calculate the expected output of peers at a segment (prev slide):  $\hat{f}(e, l, r_l, \mathbf{N}, \epsilon)$

-  $e$ : input capacity,  $l$ : layer,  $r_l$ : layer rate,  $\mathbf{N}$ : random variable denoting arrivals,  $\epsilon$ : probability of departure

- At time instance  $i$ , throughput of peers at segment  $t$  is:

$$x_{t,l}^i = \hat{f}\left( \underbrace{x_{t,l}^{i-1} \times \frac{T + T_{seed}}{T} + C_{t,l}}_I, l, r_l(t), \Lambda_t, \gamma(t) \right)$$

- Term I: input capacity to seg,  $r_l(t)$ : rate of layer  $l$  at seg  $t$ ,  $\Lambda_t$  (rnd variable): num peers at  $t$ ,  $\gamma(t)$ : departure rate at  $t$

# General Analytical Model: Step 2

## Step 2: distribution of video segments over time

$$x_{t,l}^i = \hat{f} \left( \underbrace{x_{t,l}^{i-1} \times \frac{T + T_{seed}}{T} + C_{t,l}}_I, l, r_l(t), \Lambda_t, \gamma(t) \right)$$

- $x_{t,l}^i$  converge over time;  $x_{t,l}$ : the converged value
- Expected num peers receiving layer  $l$  of segment  $t$  in the long term:  $m_{t,l} = x_{t,l} / r_l(t)$

→ Steady-state throughput:  $\sum_{t=1}^T \sum_{l=1}^L m_{t,l} \times r_l(t)$

→ Average video quality:  $\frac{1}{\sum_{t=1}^T \Lambda_t} \sum_{t=1}^T \left( \Lambda_t \sum_{l=1}^L m_{t,l} \times q_l(t) \right)$

- $\Lambda_t$ : because different num peers in different segs

# General Analytical Model: Step 2

- **Step 2: distribution of video segments over time**
  - Max peers that system can support in a flash crowd?
  - Recall: the expected num peers receiving layer  $l$  of segment  $t$  in the long term:  $m_{t,l} = x_{t,l} / r_l(t)$
  - Max of number of supportable peer =  
number of minimal substreams (base layers) that can  
be served using the expected capacity of the system =  
[total capacity – capacity needed for minimally serving  
current peers] / bitrate of base layer of first few segs =

$$\frac{1}{r_1(1)} \left( C + \sum_{t=1}^T \sum_{l=1}^L m_{t,l} \times r_l(t) - \sum_{t=1}^T \Lambda_t \times r_1(t) \right)$$

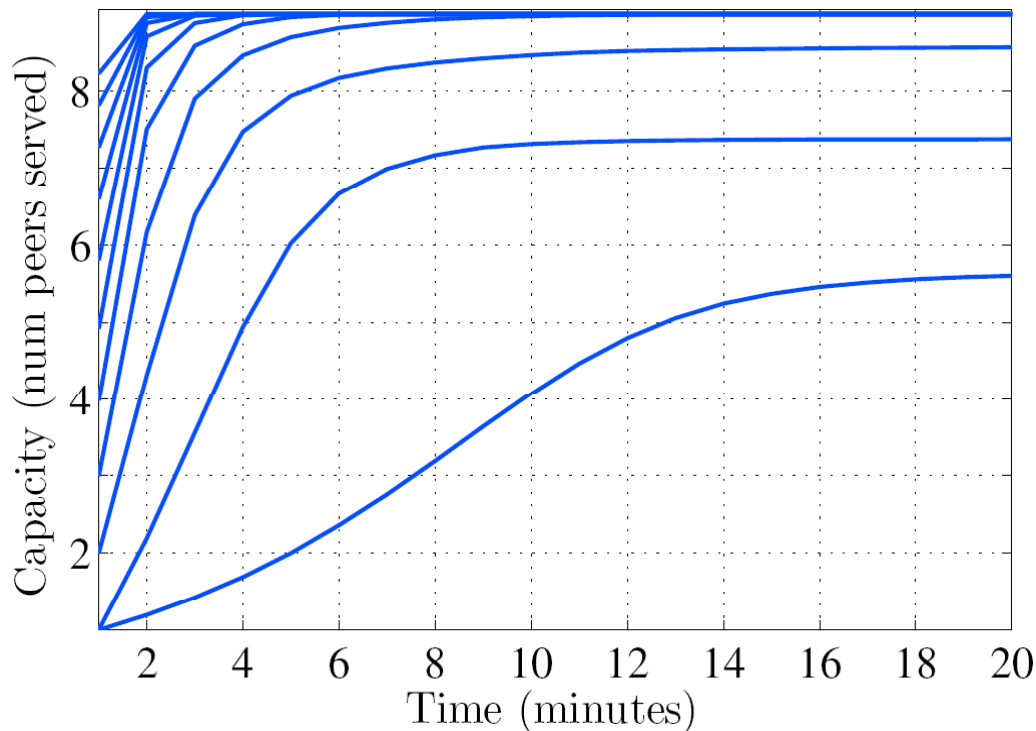
# Analysis of Example System: Procedure

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- Employing the general analytical model to analyze a sample P2P streaming system
  - Video streams:  $L$  layers, each at rate  $r$  kbps
  - Download bandwidths  $\sim U(0, M)$
  - Upload bandwidth =  $\alpha \times$  download bandwidth
  - Peer arrivals: Poisson distribution with arrival rate  $\lambda$
  - Peer departures: given probabilities  $\gamma_t$  at each seg  $t$
  - Auxiliary random variables, intermediary functions  $f_{opt | rnd | all}(\cdot)$ , and finally,  $m_{t,l}$  values calculated

# Analysis of an Example System: Results

## ■ Distribution of one layer of one video segment

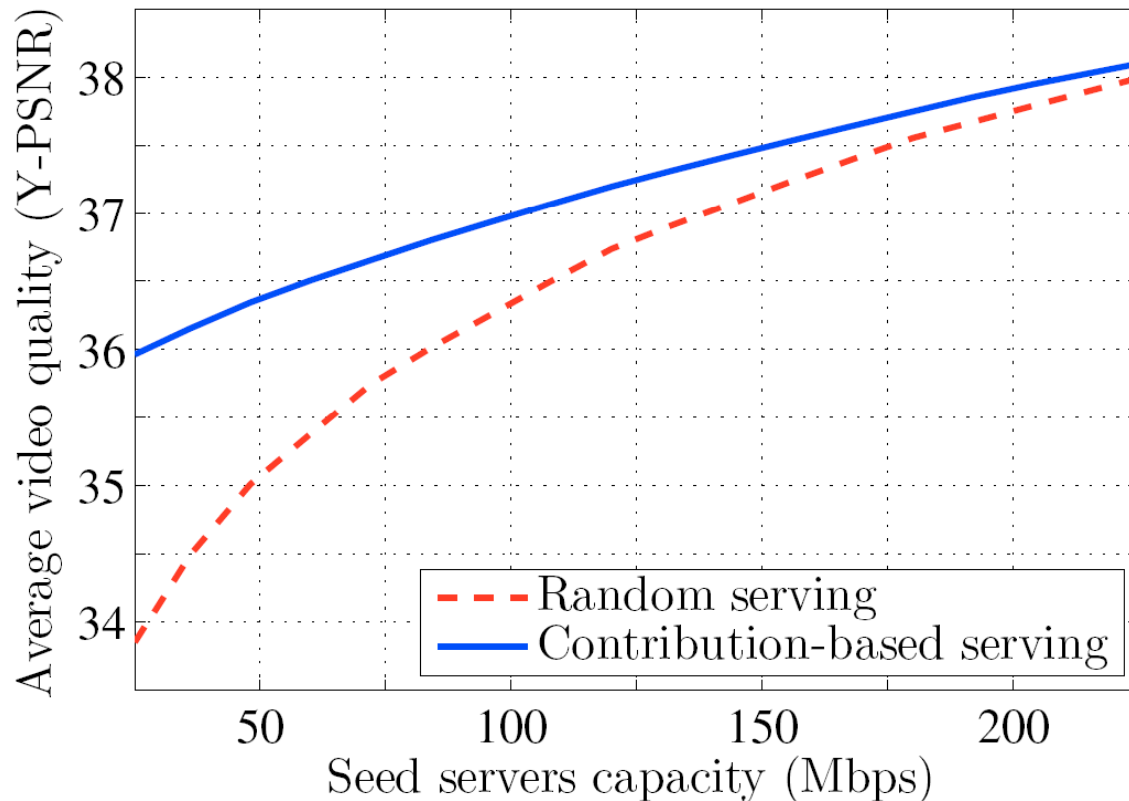


- 1-hour video, 1-min segments, 10 layers at rate 200 kbps each, max download bw: 10 Mbps, upload bw: 20% of download, arrival rate: 10 peers per minute, 25% of peers eventually leave

**Num peers receiving layer 5. Lowest curve: seed server capacity of 1 layer rate.**

# Analysis of an Example System: Results

- **Distribution of the complete video**



Average video quality delivered to peers for different seeding capacities

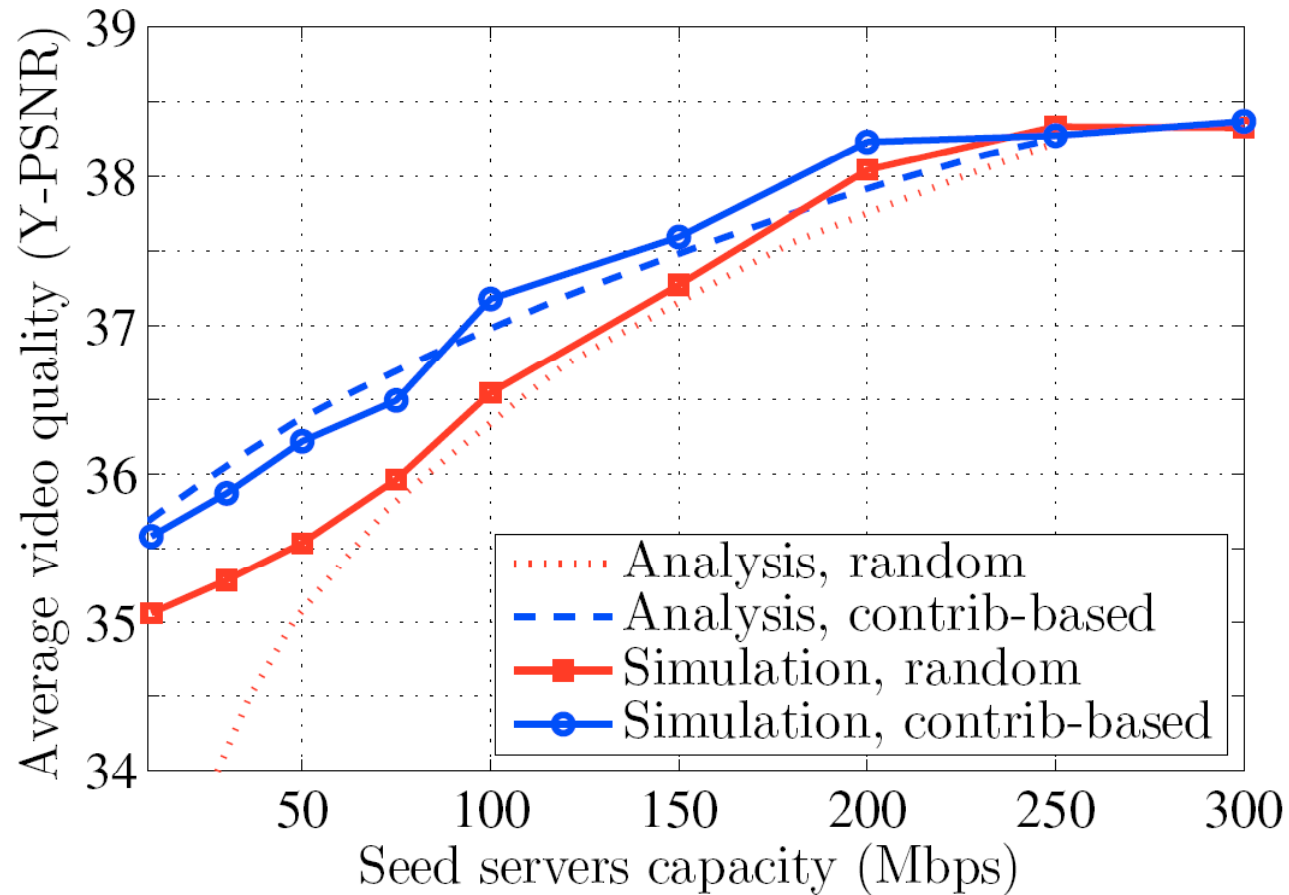
# Validation of the Analysis

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## ■ Simulation setup

- Peers and video characteristics
  - On average 500—600 peers in the network
- Event-driven simulation, 10-second time step
- Peers disobey some of our model assumptions and have other dynamics; to simulate realistic settings

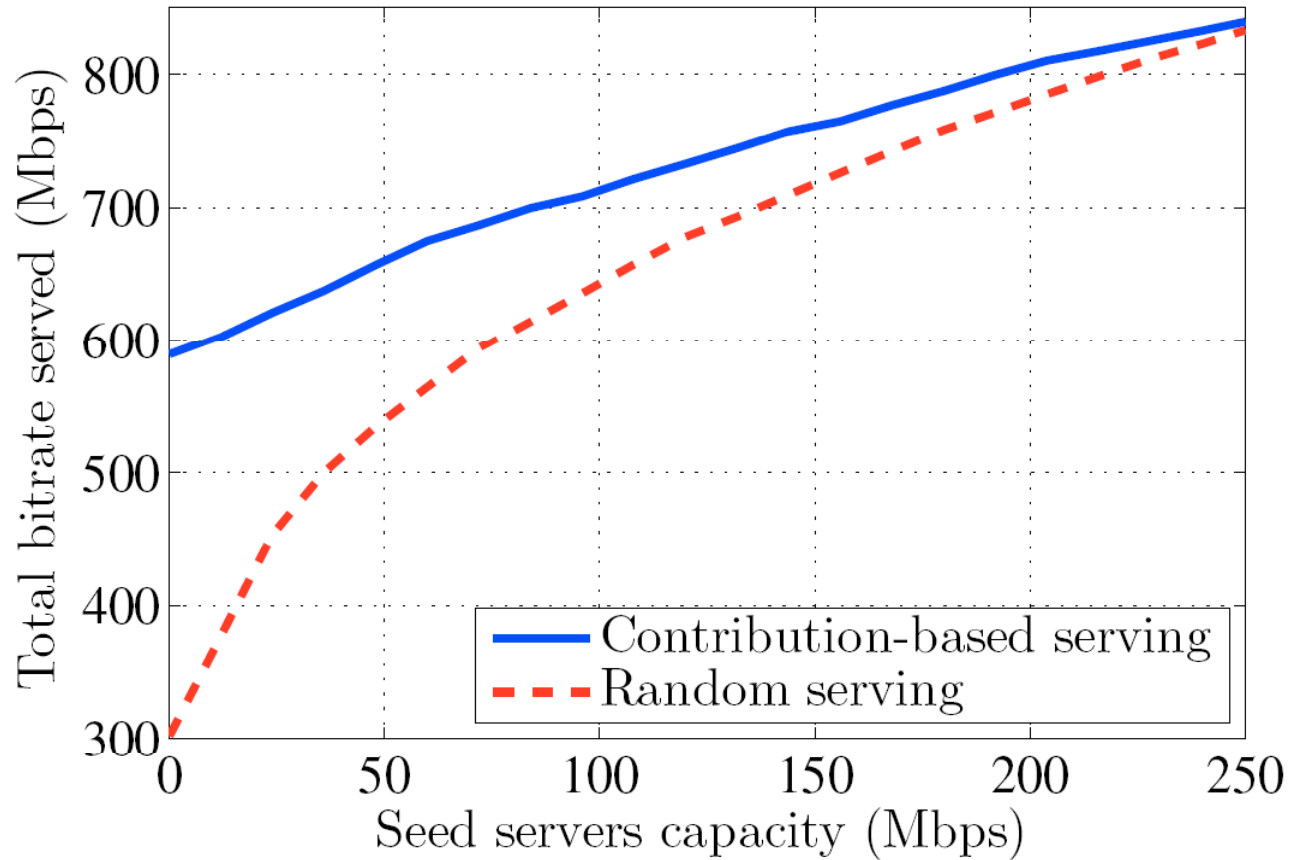
# Validation of the Analysis: Results



**Theoretical results are close to simulation results**

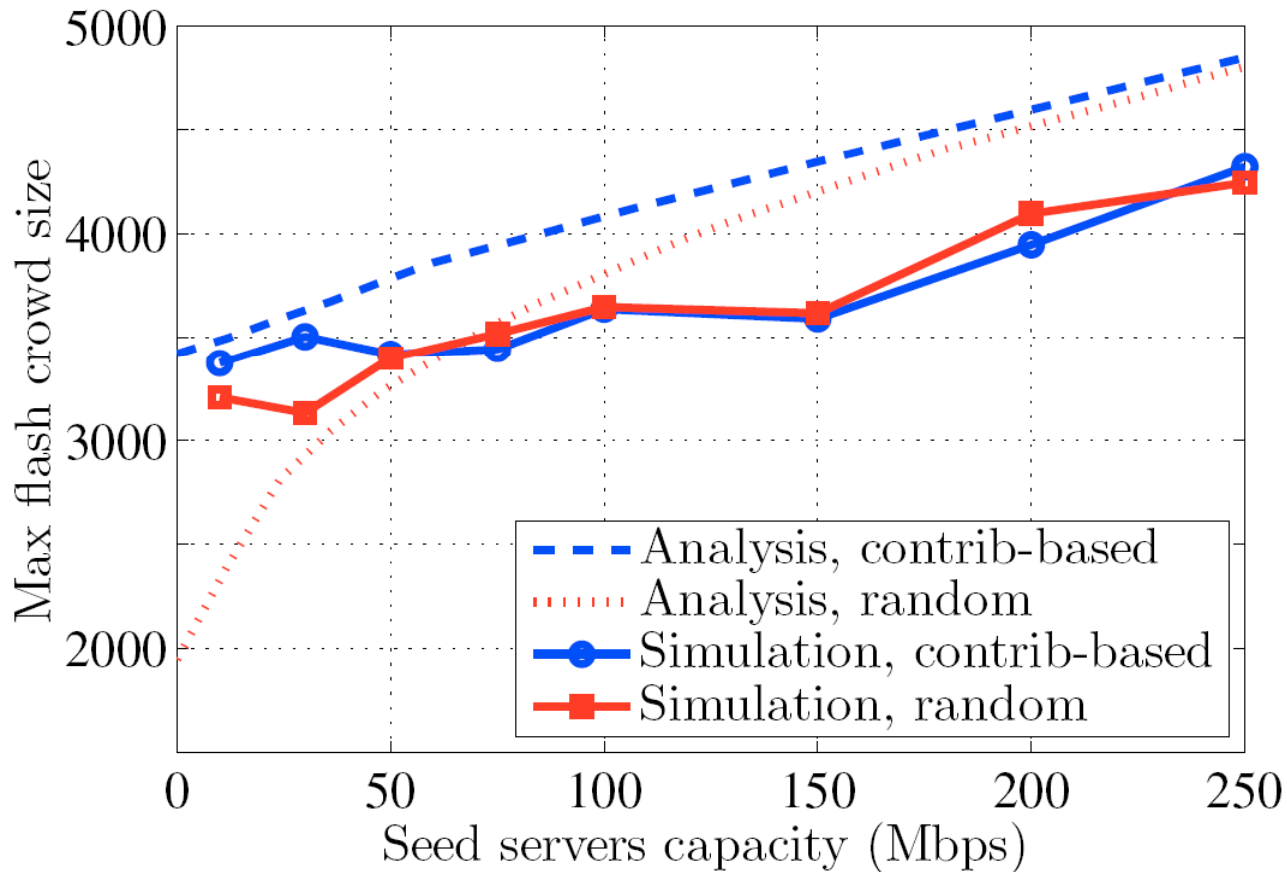


# Analyzing Seed Server Allocation Methods



**Expected throughput of the system in the long term**

# Analyzing Flash Crowd Scenarios



**Max number of peers that the system can serve (with minimal substream) in a flash crowd**

# Conclusions

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- **Theoretical analysis of P2P/peer-assisted streaming systems with scalable videos**
  - **The long-term throughput and delivered video quality**
    - **The cost-benefit tradeoff for provisioning a higher server capacity to have a desired throughput/quality**
  - **Max number of peers that the system can support in a flash crowd**
  - **Validated through simulations; results confirm the accuracy**

# Thank You!

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## Questions??

- **More info at:**

<http://nsl.cs.sfu.ca/>

# References

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