Massivizing Online Games: Distributed Computing Challenges and High Quality Time





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Technion, Haifa, IL. May 2013



Lectures at the Technion Computer Engineering Center (TCE), Haifa, IL		
IaaS Cloud Benchmarking	May 7	10am Taub 337
Massivizing Online Social Games	May 9	
Scheduling in IaaS Clouds	May 27	
Gamification in Higher Education	June 5 (HUJI)	
A TU Delft perspective on Big Data Processing and Preservation	June 6	
Grateful to Orna Agmon Ben-Yehuda, Ass	af Schuster,	Isaac Keslassy.
May 9, 2013	and Duth Be	
AISU LIIAIIKIUI LU DEIIA KULIIAII AIIU KULII DUILEITUNVersity of Technology		

The Parallel and Distributed Systems Group at TU Delft



Alexandru Iosup

Grids/Clouds P2P systems **Big Data**

Dick Epema

Grids/Clouds



Ana Lucia Varbanescu

HPC systems

Multi-cores

Big Data e-Science

tribler



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HPC systems Multi-cores P2P systems



Johan Pouwelse

P2P systems File-sharing Video-on-demand

GRENCHMARK

3

Online gaming

P2P systems Video-on-demand e-Science

Home page

www.pds.ewi.tudelft.nl

Publications

see PDS publication database at publications.st.ewi.tudelft.nl

August 31, 2011



(TU) Delft – the Netherlands – Europe



Why Social Gaming?





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What is This Talk About?



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Massivizing Social Gaming = Rich Challenge (of Distributed Computing)

Online Gaming used to be art, may now be computing

Online Gaming used to be multimedia, is now DC

Online Gaming used to be networking, is now all DC

Online Gaming used to be v-worlds, is now many apps

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DC = Distributed Computing

What's in a name? MSG, MMOG, MMO, ... Over 250,000,000 active players

Massively Social Gaming =

(online) games with massive numbers of players (100K+), for which social interaction helps the gaming experience



I. Virtual World Sim
 Explore, do, learn,
 socialize, compete
 +

2. Game Data

Player stats and relationships, others

3. Game Content

Graphics, maps, puzzles, quests, culture

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Zynga, an Amazon WS User





Source: InsideSocialGames.com

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MSGs are a Popular, Growing Market

- 25,000,000+ subscribed players (from 250,000,000+ active)
- Over 10,000 MSGs in operation
- Subscription market size \$7.5B+/year, Zynga \$600M+/year



World of Warcraft, a Traditional HPC User (since 2003)



- 10 data centers
- 13,250 server blades, 75,000+ cores
- 1.3PB storage
- 68 sysadmins (1/1,000 cores)



http://www.datacenterknowledge.com/archives/2009/11/25/wows-back-end-10-data-centers-75000-cores/

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Bungie, Computing then Serving 1.4PB/yr.



 stoshyy - JDDD
 ODST
 Image: Composition of the state of the state

- Halo 3 is one of the many successful games
- Halo 3 players get, in 1.4PB
 - Detailed player profiles
 - Detailed usage stats
 - Ranking
- CERN produces ~15PB/year (10x larger)
 - (Not) faster than the speed of light, the Higgs boson (?)

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Agenda

1. What's in a Name?

2. Three Current Challenges

- 1. Platform Scalability Challenge
- 2. Gaming Analytics Challenge
- 3. Content Generation Challenge
- 3. The Next Five Years
- 4. Conclusion



Cloudification: PaaS for MSGs

(Platform Challenge)

Build MSG platform that uses (mostly) cloud resources

- Close to players
- No upfront costs, no maintenance
- Compute platforms: multi-cores, GPUs, clusters, all-in-one!
- Performance guarantees
- Hybrid deployment model
- Code for various compute platforms—platform profiling
- Load prediction miscalculation costs real money
- What are the services?
- Vendor lock-in?
- My data

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Windows⁻Azure⁻

Mobile Social Gaming and the SuperServer

(Platform Challenge) **Support MSGs on mobile devices**

- Mobiles everywhere (2bn+ users)
- Gaming industry for mobiles is new **Growing Market**
- SuperServer to generate content for low-capability devices?
- Battery for 3D/Networked games?
- Where is my server? (Ad-hoc mobile gaming networks?)
- Security, cheat-prevention



US Mobile Gaming Revenues, by Segment, 2009-2014 millions and CAGR



Social Everything! So Analytics

- Social Network=undirected graph, relationship=edge
- Community = sub-graph, density of edges between its nodes higher than density of edges outside sub-graph

(Analytics Challenge) Improve gaming experience

- Ranking / Rating
- Matchmaking / Recommendations
- Play Style/Tutoring

Self-Organizing Gaming Communities

• Player Behavior





Content, Content, Content

(Content Challenge)

Produce and distribute content for 1BN people

- Game Analytics → Game statistic
- Crowdsourcing
- Storification
- Auto-generated game content
- Adaptive game content
- Content distribution/ Streaming content



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Impact of Platform on Game-Play

Responsive game-play

Unresponsive game-play





@large Research Challenge: V-World Platform for MMOGs

Scaling quickly to millions of players

- 1M in 4 days, 10M in 2 months
- Up-front and operational costs
- Performance, Scalability, & Cost





Rich Research Opportunities: How to Build the Core?

- Application models
- Deployment/Hosting models
- Scheduling
- Performance engineering
- Reliability engineering
- Scalability and Elasticity
- Offloading
- Etc. (including Usability, Security, Utility Models, and Programming Models)





Platform Scalability Challenge

1. Introduction

Online Game Types Game Models

- 2. (Scheduling) Hosting Models
 - 1. Cluster + Parallelization
 - 2. Multi-cluster + Sharding
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- 3. (Naming) Locating Servers
 - 1. Central directory
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- 1. (Eventual) Dead-reckoning
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Online Game Types

- MMO Role Playing Games (MMORPG)
 - Adventure: Runscape, World of Warcraft
 - Thousands of players sharing one persistent game session in a huge game world
 - Latency <1s acceptable
- First Person Shooter (FPS)
 - Action games
 - Counter Strike Source, Call of Duty, ...
 - Max. 64 players in one session (minutes or hours)
 - Latency <100ms needed

• Real-Time Strategy (RTS)

- Economic and battle strategy games
- DotA 2, Starcraft
- Latency <350ms needed







Computational Model for the Server



Game-world update Interaction computation

Entity states update

- Single sequential loop
- 3 steps in each loop:
 - 1. Game-world state update
 - 2. Entity interaction computation (dominant for MSGs)
 - 3. Entity state updates
- Load generated by (2) non-deterministic ← human factor



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Model for Entity Interaction Computation

- Player to Player/Player to Environment interaction
 - Low interaction: O(n), e.g. RTS
 - Medium interaction: O(n·log(n)) RPG quest maps
 - High interaction: O(n²) RPG war maps, FPS
 - Very High interaction: O(n² · log(n))/O(n³), unittarget matching, team path-finding, maxflow alg., ...









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Online games hosting model

- Generic Online Games (non-MM)
 - Static: dedicated isolated single servers
- MMOGs
 - Static: dedicated clusters using parallelization techniques
- Problems with these approaches

 Large amount of over-provisioning
 Non-efficient coverage of the world for the provided service



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Game parallelization models



Models:

- Zoning: huge game-world division into geographical sub-zones – each zone is handled by different machines
- Mirroring: the same game-world handled by different machines, each one handling a subset of the contained entities (synchronized states)
- **Instancing/sharding**: multiple instances of the same zone with independent states. (World of Warcraft, Runescape,..)





• Main advantages:

- 1. Significantly lower over-provisioning
- 2. Efficient coverage of the world is possible



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VENI – @larGe: Massivizing Online Games using Cloud Computing

Dynamic MSG Ecosystem: Model

Game operators

- Past player activity/business model →
 Predicted load → requests
- Data centers
 - Local time-space renting policy → offers
 - Time-Space renting policy, e.g., 1 node-hour
- Resource allocation: central request-offer matching
- Rules for ranking request-offer match: The offer size and type vs. the request, The geographical proximity offer-request, etc.



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Experimental Setup [1/3] Discrete-Event Simulator

- Input
 - Traces from RuneScape, a real top-5 MMOG
 - 7 countries, 3 continents
 - More than 130 game worlds
 - Consisting of
 - Geographical location
 - Number of clients
 - Over 10,000 samples at 2 min. interval, 2 weeks
- Output (for every time-step)
 - Resource allocation decisions
 - Resource allocation
 - Performance metrics



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Experimental Setup [3/3] Performance Metrics

- Resource over-provisioning [%]
 - The wasted resources vs. optimal provisioning at each simulation time step for all utilized machines (cumulative)
- Resource under-provisioning [%]
 - The amount of resources needed but not allocated, for all machines (computed individually)
- Significant under-provisioning events (count)
 - Count of events: resource under-provisioning is >1%, for a period of 2 minutes → people leave



Resource Provisioning and Allocation Static vs. Dynamic Provisioning



Also Studied

- Via real game measurements
 - Interactivity model (short-term msmt.)
 - Effects of underperforming platform (long-term msmt.)
- Via prototype implementation
 - Match model-reality [TPDS'11]
- Via simulation
 - Impact of virtualization [NetGames'11][IJAMC'11]
 - Economic models [CAC'13]



Remaining Challenge in Perf. Eng.: To the Real IaaS Cloud

VS



http://www.flickr.com/photos/dimitrisotiropoulos/4204766418/

- "The path to abundance"
- On-demand capacity
- Cheap for short-term tasks
- Great for web apps (EIP, web crawl, DB ops, I/O)



Tropical Cyclone Nargis (NASA, ISSS, 04/29/08)

- "The killer cyclone"
- (1) Not so great performance for scientific applications (compute- or data-intensive)
- (2) Unstable performance
- 1- Iosup et al., Performance Analysis of Cloud Computing Services for Many Tasks Scientific Computing, IEEE TPDS, 2011,

http://www.st.ewi.tudelft.nl/~iosup/cloud-perf10tpds_in-print.pdf

2- Iosup et al., On the Performance Variability of Production Cloud Services, CCGrid 2011, pds.twi.tudelft.nl/reports/2010/PDS-2010-002.pdf
Remaining Challenge in Scheduling: Provisioning and Allocation *Policies*

Provisioning

Allocation



An Example: Portfolio Scheduling for Online Gaming (also for Scientific Workloads)

- **CoH** = <u>C</u>loud-based, <u>o</u>nline, <u>Hybrid</u> scheduling
 - Intuition: keep rental cost low by finding good mix of machine configurations and billing options
 - Main idea: portfolio scheduler = run both solver of an Integer Programming Problem and various heuristics, then ⁷⁰⁰⁰ pick best schedule at deadline 6000
 - Additional feature: Can use reserved cloud instances
- 5000 Promising early results, for **€** 4000 0005 Cost **Gaming** (and scientific) workloads FCFS-CFH #jobs |average runtime [s] Trace 2000 Grid5000 200,4502728CoH 1000 LCG 188,041 8971CoH-oneType DotaLicious 109,251 2231CoH-R Heterogeneous Shen, Deng, Iosup, Epema, Scheduling Jobs in the Cloud 38 Using On-demand and Reserved Instances, EuroPar'13 (sub)

Dotalicious

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Remaining Challenge in Deployment: Zynga zCloud: Hybrid Self-Hosted/EC2

- After Zynga had large scale
- More efficient self-hosted servers
 - Run at high utilization

Time ->





May 9, 2013



11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26

"Cloud Gaming" = Interactive Gaming Videos from the Cloud

Gaikai, OnLive—servers in own data centers, HDTV
 720p
 This is only one type of cloud offloading





An Offloading Use Case: the OpenTTD Client (Early Work)



Game Parameters:

- map size
- number of players
- number of cities
- number of resources
- animations on/off





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"Cloud Gaming" = Interactive Gaming Videos from the Cloud [2/3]

- Measurement approach
 - Reverse engineer calls
 - Inject measurement probes into library calls
 - Smart detection of t3, up to and including Decoding



"Cloud Gaming" = Interactive Gaming Videos from the Cloud [3/3]

- Games used in experiments
 - Batman (action-adventure)
 - Warhammer 40k: DOW (RTS)
 - FEAR (FPS)
- Measurement results
 - OnLive better processing delay
 - FPS FEAR supported very well
 by OnLive
 - OnLive supports well Batman&DOW
 - StreamMyGame: 400-500ms delay
 - Similar playout delays

Kuan-Ta Chen et al., Measuring the latency of cloud gaming systems. ACM Multimedia 2011: 1269-1272.



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Other Distributed Systems Issues Consistency

- Eventual Consistency or Inconsistency
 - Dead-reckoning: eventual consistency with low message overhead
 - Smarter extensions



Remember RTS Games?

- Players control tens up to hundreds of units.
- Players need to take decisions in real-time.



Other Distributed Systems Issues Consistency: 1.5k Archers on 28.8k Line [1/3]

- Age of Empires [Bettner & Terrano GDC01]
- Problem: Too many players/units to update at each click
 New Approach: Simultaneous simulations next render



Source: <u>http://www.gamasutra.com/view/feature/131503/1500 archers on a 288 network .php?print=1</u>. Slides source: <u>http://www.cs.duke.edu/courses/spring08/cps214/lectures/lecture18.ppt</u>



Source: <u>http://www.gamasutra.com/view/feature/131503/1500 archers on a 288 network .php?print=1</u>. Slides source: <u>http://www.cs.duke.edu/courses/spring08/cps214/lectures/lecture18.ppt</u>

Other Distributed Systems Issues Consistency: 1.5k Archers on 28.8k Line [3/3]

• Approach: dynamic turn length

• Adjusts to real delays experienced by real players

Regular Net/CPU 200 ms latency 50 ms proc/render

Slow Net/Reg. CPU 1000 ms latency 50 ms proc/render

 Communications to	urn (200 msec) - sc	(200 msec) - scaled to 'round-trip ping' time estimates					
Process all messages	Frame	Frame	Frame				
50 msec	50 msec	50 msec	50 msec 20 fps				

(Communications turn (1000 msec) - scaled to 'round-trip ping' time estimates											
	Process all messages	Frame	Frame	Frame	000	Frame	Frame	Frame	Frame	Frame	Frame	
	50 msec20 frames, 50 msec each20						fps					

Reg. Net/Slow CPU 200 ms latency 100 ms proc/render

Communications turn (200 msec) - scaled to 'round-trip ping' time estimates				
Process all messages	Frame Frame - scaled to rendering speed			
100 msec	100 msec 10 fps			

• Problem: slow turns. Could we use only Area of Interest?

Source: <u>http://www.gamasutra.com/view/feature/131503/1500 archers on a 288 network .php?print=1</u>. Slides source: <u>http://www.cs.duke.edu/courses/spring08/cps214/lectures/lecture18.ppt</u>

Traditional AoI does not work

- Area of Interest (AoI) = traditional mechanism for RPG: only receive information around avatar, but...
- …In RTS, each player has tens of units under control, so much more data to be transferred
- ... In RTS, players change focus (interest) more often than in RPG and FPS, so higher management overhead



Core Idea

- Partition the game into multiple areas (rectangular)
- Each player pays attention to different areas + attention level
- Depending on attention level and machine performance, the player will receive different types of information (commands or state) about the game world
 - AoS: Area of Simulation = high-attention area, full simulation based on input commands (CPU-intensive)
 - AoU: Area of Update = low-attention area, receives state (Net)
 - NUA: No update area
- Each player can have multiple AoS and AoU



Work in Progress

Experimental results

- Simulator and prototype RTS game
- Evaluate in two Cloud platforms: EC2 and Azure
- Prototype about 20k lines of C++ code
 - Based on an open source game (~6k lines)
- Up to 200 players and **10,000 battle units**
 - State-of-the-art unplayable at 1-2,000
 - Crashes not uncommon due to CPU and Network bottlenecks
- → Using our AoS-based method can lead to lower CPU consumption than pure event-based method (RTS) and lower network consumption than pure update-based

method (RPG)

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Work in Progress

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@large: Sample Analytics Results Analysis of Meta-Gaming Network

- "When you play a number of games, not as ends unto themselves but as parts of a larger game, you are participating in a metagame." (Dr. Richard Garfield, 2000)
- XFire: since 2008 (3+ years), 500K of 20M players



@large Research Challenge: Continuous Analytics for MMOGs

Analyzing the behavior of millions of players, on-time

- Data mining, data access rights, cost v. accuracy, ...
- Reduce upfront costs
- Low response time & Scalable
- Large-scale Graph Processing



 2.5×10^{5}

2.0 x 10

How to Analyze? Ecosystems of **Big-Data Programming Models** High-Level Language SQL Meteor JAQL Hive Pig Sawzall Flume BigQuery Scope DryadLINQ AQL oarammina Model MapReduce Model PACT Dataflow Pregel Algebrix **Execution** Engine Haloop Hadoop/ Giraph Flume Dremel Tera Azure Nephele MPI/ Dryad Hyracks YARN Engine Service Data Engine Erlang Tree Engine Storage Engine **S**3 **GFS** HDFS Voldemort L CosmosFS Tera Azure Asterix Data Data **B-tree** S Store Store * Plus Zookeeper, CDN, etc.

Adapted from: Dagstuhl Seminar on Information Management in the Cloud, http://www.dagstuhl.de/program/calendar/partlist/?semnr=11321&SUOG





DotA communities





- Players are loosely organised in communities
 - Operate game servers
 - Maintain lists of tournaments and results
 - Publish statistics and rankings on websites
- Dota-League: players join a queue and matchmaking forms teams
- DotAlicious: players can choose which match/team to join



Our Datasets

- We have crawled all matches played and per match have:
 - Names of the players for each team
 - Active, start and end time
 - Game-play statistics per team
 - The team that has won
- Dota-League:
 - ~1.5M matches played between Nov'08 and Jul'11, 61K players
- DotAlicious:
 - ~0.6M matches played between Apr'10 to Feb'12, 62K players



From game instances to social ties

- We need to define how to map the relationships found in real-world matches to a **gaming graph** (nodes and links)
- We use six different mappings and various thresholds:
 - **SM:** two players occur more than *n* times in the **same match**
 - SS: two players occur more than n times on the same side
 - **OS:** two players occur more than *n* times on **opposing sides**
 - ML: two players have lost more than *n* matches together
 - MW: two players have won more than *n* matches together
 - **PP:** a directed version of the mappings above. A link exists if a player has played more than n percent of his matches together







Relationships in the gaming graph

- Players who regularly play together in DotAlicious do so in more diverse combinations than in Dota-League
- Contrary to Dota-League, DotAlicious players tend to play on the same side: playing together intensifies the social bond
- Winning together increases friendship relationships, while loosing together weakens friendship relationships
- Small clusters of friends with very strong social ties exist
- R. van de Bovenkamp, S. Shen, A. Iosup, F. A. Kuipers: Understanding and recommending play relationships in online social gaming. COMSNETS 2013: 1-10



Matchmaking application

- Replay match list, but also consider clusters in gaming graph
- Scoring methodology:
 - Points per cluster: Number of players in the match that are part of the same cluster
 - Excluding largest cluster of the network and clusters of size 1



Results matchmaking DotAlicious **Dota-League** 2.5 2.5 Original 🖊 Matchmaking Random \sim Original Matchmaking Random \sim \checkmark 2.0 2.0 1.5 1.5 Score Score 1.01.00.5 0.5 0.0 0.0 SS SM OS ML MW SM SS OS ML MW Can do much better than random matchmaking

Can already improve original matchmaking algorithm for all gaming graphs!



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@large: Sample Analytics Results Skill Level Distribution in RuneScape

- Runescape: 135M active accounts, 7M active (2008)
- High-scoring players: 1.8M (2007) / 3.5M (2010)



@large Research Challenge: Content Generation for MMOGs

Generating content on time for millions of players

- Player-customized: Balanced, Diverse, Fresh
- Up-front and operational costs
- Response time, Scalability, & Cost







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(Procedural) Game Content (Generation)





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The POGGI Content Generation Framework


Puzzle-Specific Considerations Generating Player-Customized Content

Puzzle difficulty

- Solution size
- Solution alternatives
- Variation of moves
- Skill moves



Player ability

- Keep population statistics and generate enough content for most likely cases
- Match player ability with puzzle difficulty
- Take into account puzzle freshness

B:Up X:Up B:Left C:Down C:Left B:Down B:Right B:Down E:Right E:Down E:Right B:Up A:Up B:Left C:Down C:Right E:Down X:Left E:Left X:Down X:Left (Best solution: 21 moves)



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Agenda

- 1. What's in a Name?
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3. The Next Five Years

1. Cloudification

- 2. Mobile Social Gaming
- 3. Social Everything!
- 4. Content, Content, Content

4. Conclusion





Cloudification: PaaS for MSGs

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- SuperServer to generate content for low-capability devices?
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- Where is my server? (Ad-hoc mobile gaming networks?)
- Security, cheat-prevention



US Mobile Gaming Revenues, by Segment, 2009-2014 millions and CAGR



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Social Everything!

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(Analytics Challenge) Improve gaming experience

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Self-Organizing Gaming Communities

• Player Behavior





Content, Content, Content

(Content Challenge)

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Massivizing Online Gaming

- Million-user, multi-bn market
- V-World, Content, Analytics

Current Technology

- Upfront payment
- Cost and scalability problems
- Makes players unhappy

```
Publications Gaming and Clouds
2008: ACM SC
2009: ROIA, CCGrid, NetGames,
EuroPar (Best Paper Award), ...
2010: IEEE TPDS, Elsevier CCPE
2011: Book Chapter CAMEO, IEEE
TPDS, IJAMC
2012: IPDPS, CCGrid, ...
Graduation (Forecast)
2012-14: 3PhD, 6Msc, 6BSc
```

Summary

@large: Our Vision

- DC/Cloud has to help
- Economy of scale with clouds

@large: Ongoing Work

- Content: POGGI Framework
- Platform: edutain@grid
- Analytics: CAMEO Framework

@large: The Future

- Happy players
- Happy cloud operators

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Thank you for your attention! Questions? Suggestions? Observations?

More Info:

- http://www.st.ewi.tudelft.nl/~iosup/research.html
- http://www.st.ewi.tudelft.nl/~iosup/research gaming.html
- http://www.st.ewi.tudelft.nl/~iosup/research_cloud.html

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http://www.pds.ewi.tudelft.nl/~iosup/ (or google "iosup") Parallel and Distributed Systems Group Delft University of Technology

Massivizing Social Games: High Performance Computing and High Quality Time – A. Iosup



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Do not hesitate to

contact me...

Massivizing Social Games: Distributed Computing Challenges and High Quality Time – A. Iosup



Platform Scalability Challenge

1. Introduction

- 1. Online Game Types
- 2. Game Models
- 2. (Scheduling) Hosting Models
 - 1. Cluster + Parallelization
 - 2. Multi-cluster + Sharding
 - 3. Cloud-based + Sharding
 - 4. The SuperServer (video)

3. (Naming) Locating Servers

- **1. Central directory**
- 2. Peer-to-Peer

4. Consistency

- 1. (Eventual) Dead-reckoning
- 2. (Continuous) Simultaneous Sim
- 3. (Continuous) + Pipelining
- 4. (Continuous) + Dynamic Tick
- 5. (Continuous/Eventual) Area of Simulation vs Interest
- 5. Other issues



Other Distributed Systems Issues Naming (Locating Servers)

- Master directory
 - Web page with hundreds of servers (Minecraft)
 - TCP/IP-based protocol to get information from master server



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Other Distributed Systems Issues

- Operation of replicated servers: performance guarantees
- Operation with slow user clients/networks
- Persistent worlds
- Content distribution
- The whole CAP spectrum
 - Consistency
 - Availability
 - Partition-tolerance



http://www.popscreen.com/v/6wEHS/Minecraft-Epic-Fail-Creepe



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Impact on Game Experience

Responsive game

Unresponsive game



May 9, 2013

[Source: Nae, Iosup, and Prodan, ACM SC 2008 and IEEE TPDS 2011]



Experimental Setup [2/3] Environment

- 1 game operator
- 17 data centers
- 11 data center time-space renting policies

Location		Data	Machines (total)	
Continent	Country	Centers	wachines (total)	
Europe	Finland	2	8 machines	
	Sweden	2	8 machines	
	U.K.	2	20 machines	
	Netherlands	2	15 machines	
	U.S. (West)	2	35 machines	
	Canada (West)	1	15 machines	
North America	U.S. (Central)	1	15 machines	
	U.S. (East)	2	32 machines	
	Canada (East)	1	10 machines	
Australia	Australia	2	8 machines	



[Source: Nae, Iosup, and Prodan, ACM SC 2008]

Impact of Load Prediction Accuracy Q: How does the prediction accuracy impact resource provisioning? A: Good prediction matters.



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Impact of Interaction Compute-Intensiveness

Q: How are different MMOG types handled under dynamic resource provisioning?

(Interaction models Low ~ O(n), Medium, High, Very High ~ O(n³))



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Latency Tolerance: From None to High

Q: What is the impact of latency tolerance on hosting?



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@large: Sample Analytics Results Activity and Social Network

Bridge Base Online (BBO): 1M+ players, top free site

250000

- Dataset: **100K players**
 - 9K group
 - Social relationships from bridge pairing
- Large (~10K) online social groups can coordinate



community builder, community member, random player, faithful player

M. Balint, V. Posea, A. Dimitriu, and A. Iosup, An Analysis of Social Gaming Networks in Online and Face to Face Bridge Communities, LSAP 2011.



50000

Resource Provisioning and Allocation Compound Metrics



- Trade-off Utility-Cost still needs investigation
- Performance and Cost are not both improved by the policies we have studied

Massivizing Socia	Villegas, Antoniou, Sadjadi, Iosup. An Analysis of
	Provisioning and Allocation Policies for Infrastructure-
	as-a-Service Clouds, CCGrid, 2012.



- Performance Evaluation of Four Commercial Clouds
 - Amazon EC2, GoGrid, Elastic Hosts, Mosso
 - Resource acquisition
 - Single- and Multi-Instance benchmarking
- Low compute and networking performance¹
- Performance variability over time²



 1- Iosup et al., Performance Analysis of Cloud Computing Services for Many Tasks Scientific Computing, IEEE TPDS, 2011, <u>http://www.st.ewi.tudelft.nl/~iosup/cloud-perf10tpds_in-print.pdf</u>
 2- Iosup et al., On the Performance Variability of Production Cloud Services, CCGrid 2011, <u>pds.twi.tudelft.nl/reports/2010/PDS-2010-002.pdf</u>

Multi-Resource Provisioning/Release



Time for *multi*-resource increases with number of resources

Massivizing Social Games: High Performance Computing and High Quality Time – A. Iosup IOSUP et al., Performance Analysis of Cloud Computing Services for Many Tasks Scientific Computing, (IEEE TPDS 2011).

GAE Dataset: Run Service



- Fibonacci [ms]: Time it takes to calculate the 27th Fibonacci number
- Highly variable performance until September
- Last three months have stable performance (low IQR and range)

Massivizing Social <u>Games: High Performance Computing and High Quality Time – A. Iosup</u> IOSUP, Yigitbasi, Epema. On the Performance Variability of Production Cloud Services, (IEEE CCgrid 2011).

Online Scheduling + Optimization ExPERT



 $ExPERT recommended \equiv$

 $(N = 3, T = T_{ur}, D = 2T_{ur}, M_r = 0.02)$, in words: Send N = 3 instances to the unreliable pool during the tail phase, each timed out after twice the average task time $(D = 2T_{ur})$. Send the next instance after the average task time passes $(T = T_{ur})$. Use only one $(\sharp ur = 50, 50 \times M_r = 1)$ reliable machine at a time.

O. Agmon Ben-Yehuda, A. Schuster, A. Sharov, M. Silberstein, and A. Iosup, EXPERT: Pareto-Efficient Task Replication on Grids and a Cloud, IPDPS'12.

Performance Metrics





- Makespan very similar
- Very different job slowdown

Massivizing Socia Villegas, Antoniou, Sadjadi, Iosup. An Analysis of Provisioning and Allocation Policies for Infrastructureas-a-Service Clouds, (submitted). PDS Tech.Rep.2011-009



- Very different results between actual and charged
 - Cloud charging function an important selection criterion
- All policies better than Startup in actual cost
- Policies much better/worse than Startup in charged cost

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Massivizing Socia	Villegas, Antoniou, Sadjadi, Iosup. An Analysis of
	Provisioning and Allocation Policies for Infrastructure-
	as-a-Service Clouds, (submitted). PDS Tech.Rep.2011-009

Single Resource Provisioning/Release





- Time depends on instance type
- Boot time non-negligible

Massivizing Social Games: High Performance Computing and High Quality Time – A. Iosup IOSUP et al., Performance Analysis of Cloud Computing Services for Many Tasks Scientific Computing, (IEEE TPDS 2011).

CPU Performance of Single Resource

- ECU definition: "a 1.1 GHz 2007 Opteron" ~ 4 flops per cycle at full pipeline, which means at peak performance one ECU equals 4.4 gigaflops per second (GFLOPS)
- Real performance
 0.6..0.1 GFLOPS =
 ~1/4..1/7 theoretical peak



Massivizing Social Games: High Performance Computing and High Quality Time – A. Iosup IOSUP et al., Performance Analysis of Cloud Computing Services for Many Tasks Scientific Computing, (IEEE TPDS 2011).





AWS Dataset (4/4): Summary



All services exhibit time patterns in performance

- EC2: periods of special behavior
- SDB and S3: daily, monthly and yearly patterns
- SQS and FPS: periods of special behavior

Massivizing Social <u>Games: High Performance ComputiMars</u> 2013 Duality Time - A. <u>Iosup</u> IOSUP, Yigitbasi, Epema. On the Performance Variability of Production Cloud Services, (IEEE CCgrid 2011).



GAE Dataset (2/4): Datastore



- Read Latency [s]: Time it takes to read a "User Group"
- Yearly pattern from January to August
- The last four months of the year exhibit much lower IQR and range
 - More stable performance for the last five months
 - Probably due to software/infrastructure upgrades

Massivizing Social <u>Games: High Performance Computimary</u> 2013 Quality Time - A. Josup IOSUP, Yigitbasi, Epema. On the Performance Variability of Production Cloud Services, (IEEE CCgrid 2011).

GAE Dataset (3/4): Memcache



- **PUT [ms]:** Time it takes to put 1 MB of data in memcache.
- Median performance per month has an increasing trend over the first 10 months
- The last three months of the year exhibit stable performance

Massivizing Social <u>Games: High Performance ComputiMarr9</u>, <u>2013</u><u>Duality Time – A. Iosup</u> IOSUP, Yigitbasi, Epema. On the Performance Variability of Production Cloud Services, (IEEE CCgrid 2011).

GAE Dataset (4/4): Summary



- All services exhibit time patterns
- Run Service: daily patterns and periods of special behavior
- Datastore: yearly patterns and periods of special behavior
- Memcache: monthly patterns and periods of special behavior
- URL Fetch: daily and weekly patterns, and periods of special behavior

Massivizing Social <u>Games: High Performance ComputiMarr</u> 2 2013 Quality Time - A. <u>Iosup</u> IOSUP, Yigitbasi, Epema. On the Performance Variability of Production Cloud Services, (IEEE CCgrid 2011).

@large: Social Everything!

- Social Network=undirected graph, relationship=edge
- Community = sub-graph, density of edges between its nodes higher than density of edges outside sub-graph

(Analytics Challenge) Improve gaming experience

- Ranking / Rating
- Matchmaking / Recommendations
- Play Style/Tutoring

Self-Organizing Gaming Communities

• Player Behavior





The Game Trace Archive (upcoming)

- Share gaming traces and best-practices on using them
- Support simulations and real-world experiments

Name	Period	Size (GB)	Node (M)	Edge (M)	Category
KGS	2002/02-2009/03	2	0.8	27.4	Chess Game
FICS	1997/11-2011/09	168	0.4	144.2	Chess Game
BBO	2009/11-2009/12	10	3.9	12.9	Card Game
XFire	2008/05-2011/12	58	7.7	34.7	OMGN
Dota League	2006/07-2011/03	23	0.1	3.0	RTS
DotaLicious	2010/04-2012/02	1	0.1	0.6	RTS
Dota Garena	2009/09-2010/05	1	0.3	0.1	RTS
WoWAH	2006/01-2009/10	3	0.1	N/A	MMORPG


The CAMEO Framework



1. Address community needs

- Can analyze skill level, experience points, rank
- Can assess community size dynamically

2. Using on-demand technology: Cloud Comp.

• Dynamic cloud resource allocation, Elastic IP

3. Data management and storage: Cloud Comp.

• Crawl + Store data in the cloud (best performance)

4. Performance, scalability, robustness: Cloud Comp.

A. Iosup, CAMEO: Continuous Analytics for Massively Multiplayer Online Games on Cloud Resources. ROIA, Euro-Par 2009 Workshops, LNCS 6043, (2010)



The POGGI Framework



Focus on game content generation on grids

- Use existing middleware
- Control MMOG-specific workload demands and variability (soft guarantees for low response time by pre-generating content)

... but do not forget lessons on system design

• Add components for capacity planning and process monitoring

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Continuous Analytics for MMOGs

MMOG Data = raw and derivative information from the virtual world (millions of users)

Continuous Analytics for MMOGs = Analysis of MMOG data s.t.

- Data collection
- Data storage
- Data analysis
- Data presentation
- ... at MMOG rate and scale









Continuous Analysis for MMOGs Main Uses By and For Gamers

- 1. Support player communities
- 2. Understand play patterns (decide future investments)
- 3. Prevent and detect cheating or disastrous game exploits (think MMOG economy reset)
- 4. Broadcasting of gaming events
- 5. Data for advertisement companies (new revenue stream for MMOGs)



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Other Uses for MMOG Data

Social Sciences

- The emergence and performance of ad hoc groups in contemporary society
- Emergent behavior in complex systems

Economy

Contemporary economic behavior

Psychology

- Games as coping mechanism (minorities)
- Games as cure (agoraphobia)

Biology

• Disease spread models

Voorpagina Algemeen Economie Sport Internet Achterklap Utgegeven: 2 mei 2009 11:45 Lasist oewiiziidi 2 mei 2009 11:45

Last Updated: Tuesday, 21 August 2007, 00:04 GMT 01:04 UK

http://news.bbc.co.uk/2/hi/health/6951918.stm BBC Virtual game is a 'disease model'

An outbreak of a deadly disease in a virtual world can offer insights into real life epidemics, scientists suggest.

The "corrupted blood" disease spread rapidly within the popular online World of Warcraft game, killing off thousands of players in an uncontrolled plague.



Scientists believe the game error could offer a valuable insight

The infection raged, wreaking social chaos, despite quarantine measures.

The experience provides essential clues to how people behave in such crises, Lancet Infectious Diseases reports.

In the game, there was a real diversity of response from the players to the threat of infection, similar to those seen in real life.

Some acted selflessly, rushing to the aid of other characters even though that meant they risked infection themselves.

Others fled infected cities in an attempt to save themselves. 66 The players seemed to really feel they were at risk and took the threat of infection seriously

Professor Nina Fefferman, from Tufts University School of Medicine

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GAMES



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The CAMEO Framework [ROIA09] Continuous MSG Analytics on the Cloud



- Use own resources for continuous or predicted load
- Use cloud (on-demand, paid-for, guaranteed) resources for sparse or excess load
- Users (peers) may also provide service (future)

A. Iosup, CAMEO: Continuous Analytics for Massively Multiplayer Online Games on Cloud Resources. ROIA, Euro-Par 2009 Workshops, LNCS 6043, pp. 289--299. Springer, Heidelberg (2010)



CAMEO: Analytics Capabilities

1. Various pieces of information

• Skill level, experience points, rank

2. Single and Multi-snapshot analysis

3. Analysis functions already implemented

- Ranking by one or more pieces of information
- Community statistical properties for a piece of information
- Identification of Top-K players in single/multi-snapshot
- Evolution of (Top-)K players
- Evolution of average community skill
- Identification of players with special skill combos



CAMEO: Cloud Resource Management



- Snapshot = dataset for a set of players
- More machines = more snapshots per time unit

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CAMEO: Exploiting Cloud Features

Machines close(r) to server

Traffic dominated 100
by small packets (latency) 75



• Elastic IP to avoid traffic bans

(legalese: acting on behalf of real people)

A. Iosup, A. Lascateu, N. Tapus, CAMEO: Enabling Social Networks for Massively Multiplayer Online Games through Continuous Analytics and Cloud Computing, ACM NetGames 2010.



Cost of Continuous RuneScape Analytics

Billing Statement: April 1, 2009

Billing Cycle for this Report: March 1 - March 31, 2009

			Expand All Collapse All
Rate		Usage	Totals
Amazon Elastic Compute Cle View/Edit Service	oud		
Amazon EC2 running Linux/UNIX			
\$0.10 per Small Instan hour (or partial hour)	ce (m1.small) instance-	2,097 Hrs	209.70
Amazon EC2 Bandwidth			
\$0.100 per GB Internet transfer into Amazon E0	: Data Transfer - all data C2	611.005 GB	61.10
\$0.170 per GB Internet TB / month data transf	Data Transfer - first 10 er out of Amazon EC2	507.121 GB	86.21
Taxes			67.83
Charges due on April 1, 2009+			424.85

- Put a price on MMOG analytics (here, \$425/month, or less than \$0.00015/user/month)
- Trade-off accuracy vs. cost, runtime is constant

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Performance Results: Why Choosing the Cloud Matters



 Location of machines influences MMOG analytics performance (data acquisition)

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Player-Customized Content Skill Level Distribution in RuneScape

- **RuneScape**: 135M+ open accounts (world record)
- Dataset: 3M players (largest measurement, to date)
 - 1,817,211 over level 100 2.4*10⁵
 - Max skill 2,280
- Number of mid- and high-level players is significant

New Content Generation Challenge



The POGGI Content Generation Framework



Only the puzzle concept, and the instance generation and solving algorithms, are produced at development time

^{*} A. Iosup, POGGI: Puzzle-Based Online Games on Grid Infrastructures, EuroPar 2009 (Best Paper Award)



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Grid Workload Components



Workflow Engines:

Performance vs. Resource Consumption

Middleware	MS [s]
DAGMan	$1,327 \pm 138$
Karajan	$1,111 \pm 154$

Karajan performs better than DAGMan, but runs quickly out of resources.



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