

Beyond Spreadsheets:

How to generate operational impact
with analytics

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about me ...

- » Studied **Media Management** with an empirical focus
- » 2010-2013 Games Analyst, Lead Games Analyst and Head of Analytics at **Bigpoint** GmbH
- » Since 2013 Head of Analytics at **InnoGames** GmbH

about InnoGames ...

- » Started 2003 as a Hobby-project from our Managing Directors, today the classic Tribal Wars is still growing after 11 years
- » F2P-Business-model, started on browser, transition to **cross-platform** developer & publisher
- » **330 employees** from 25 nations, 400 until the end of the year
- » 130 Mio. registered players, ~**70 million €** in revenues in 2013 (and growing)







spreadsheets are not all that bad ...

2 examples of valuable standard-reporting

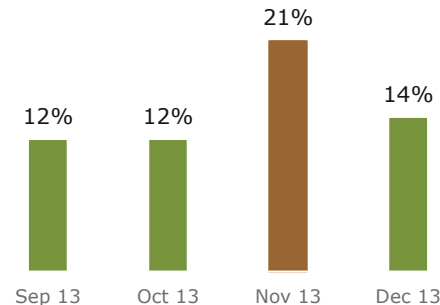
1.

tutorial-conversion-funnel reporting

→ 75% higher drop out for FoE in first quest due to synced bug



drop out for first tutorial quest



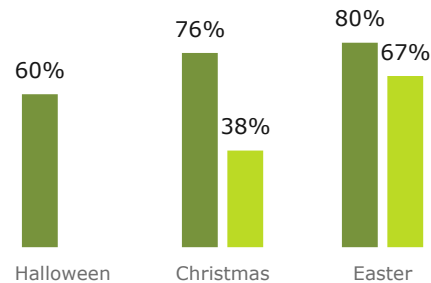
2.

event evaluations

→ constant improvement of participation rates due to awareness and detail-optimization



participation rates for major Grepolis-events

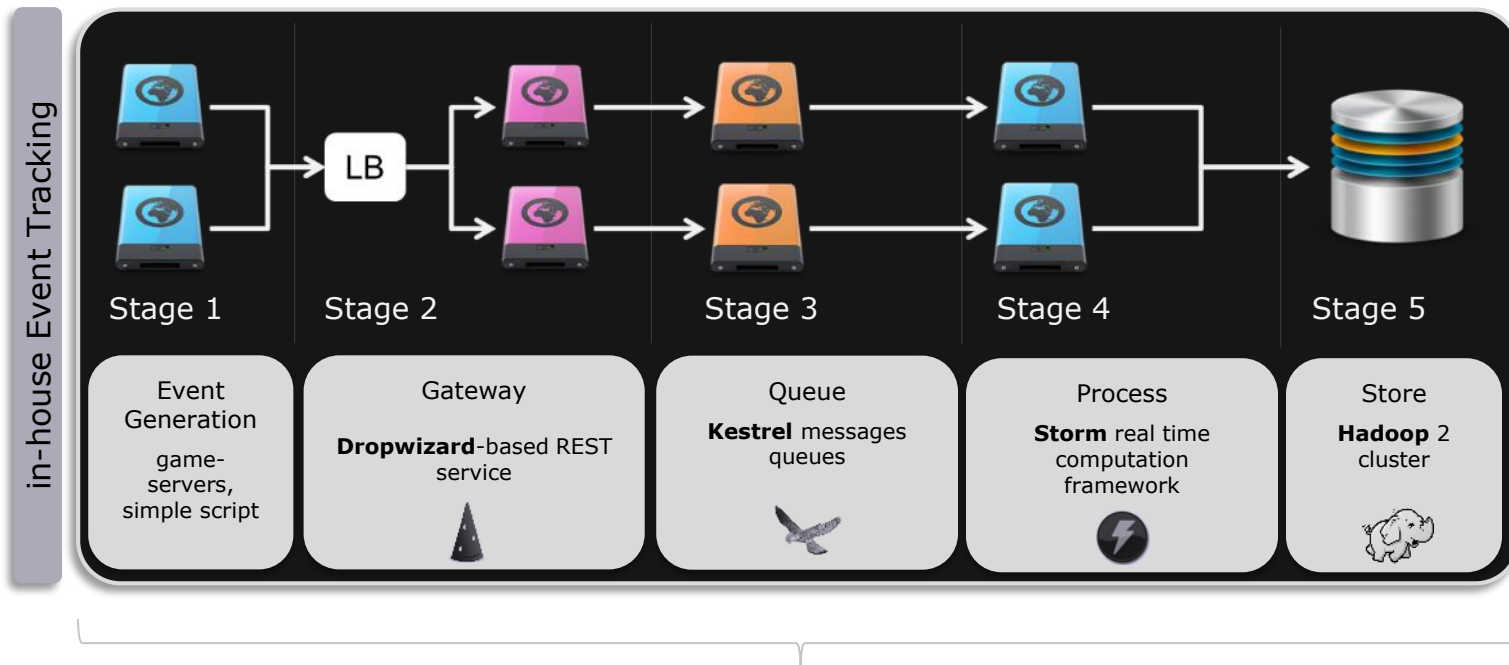


■ Browser ■ Mobile



still, there's more data ...

backend



analytics
data
mining
team



all stages scale out and use open-source-software





... too good just for fun facts

60 GB new event-data every day

saved on a Hadoop-cluster with currently **44** servers

> 400 M logged events every day

every day ...

1,1 Mio. hours of InnoGames are played all around the globe

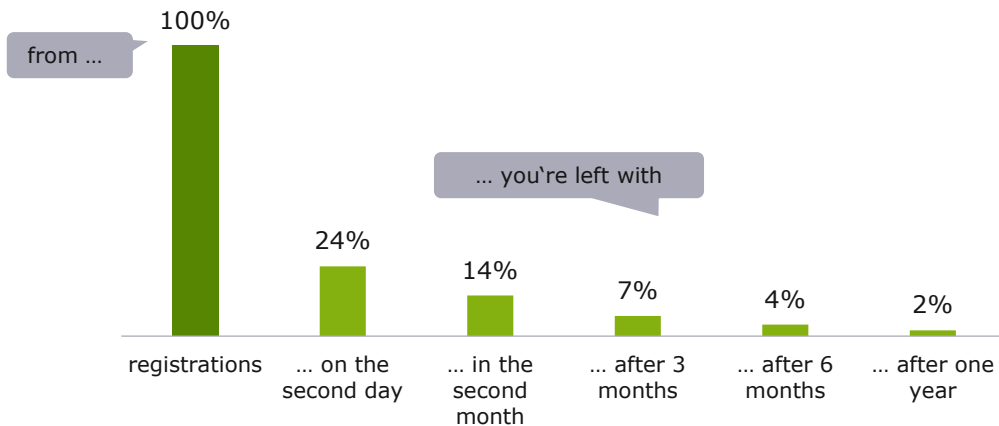
10 Mio. neighbors are visited in Forge of Empires

74 Mio. units are recruited in Grepolis

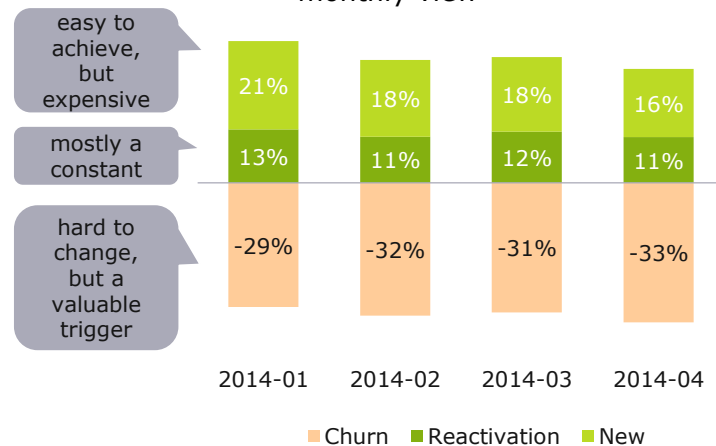


churn: the hard truth about free-to-play

spreading loss is high: cohort-view



a stable user-basis requires constant input: monthly view



churn reasons

Naturally, most churn happens whatever you do ...



was conquered/attacked too often, alliance disbanded

boredom, perceived unfairness, not enough freedom, too simple, too complex



... but there is enough potential where incentives & communication might help



first steps in prediction modelling

GREPOLIS



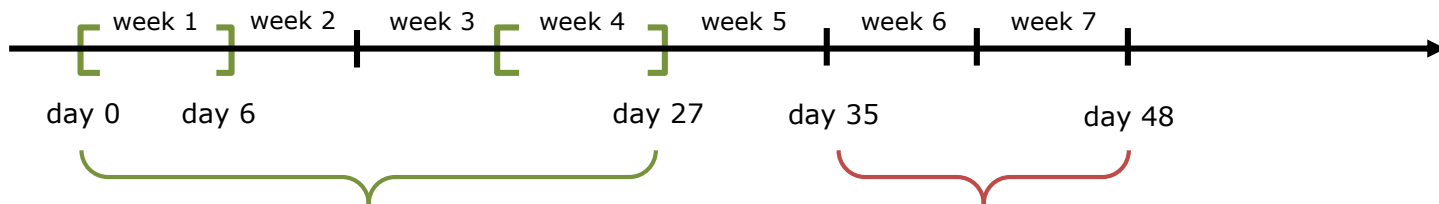
predict churn of mid-game players at a time where they are still active

43 defined parameters, e.g. playtime, deff-battles, off-battles, completed quests ...

REGISTRATION

ACTIVITY

PREDICTION



Prediction input is based on all events within the first 4 weeks after registration

Goal: Predicting if player will churn or stay active in week 6 + 7

decision tree models

Classification of objects based on decision-rules until 'optimal' classification is reached

R party-package

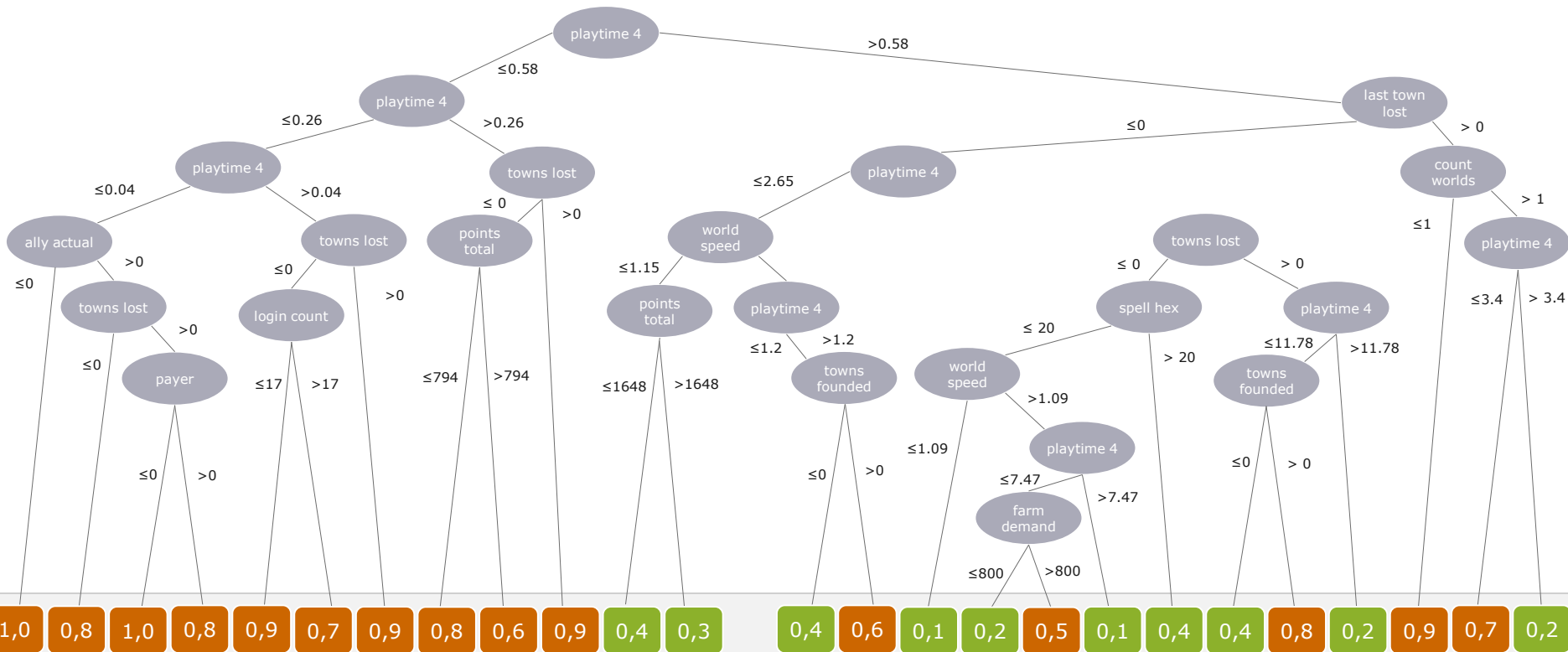
works with all types of data, controls for overfitting

benchmark of different parameter-sets to find a sweet-spot between active- and churn-prediction



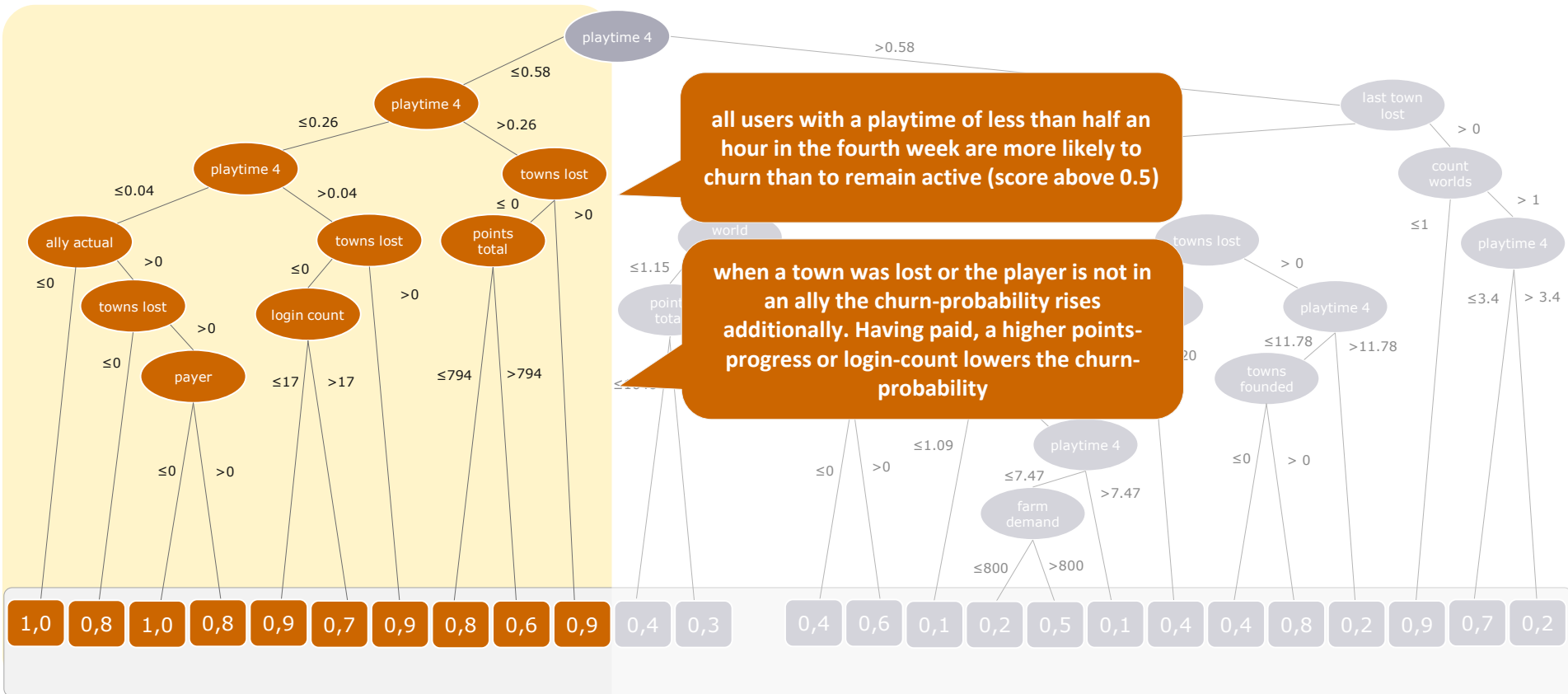
actually quite simple: a decision tree model

GREPOLIS



Probability of churning: Below 0,5 = user is scored as active, above 0,5 = user is scored as churned

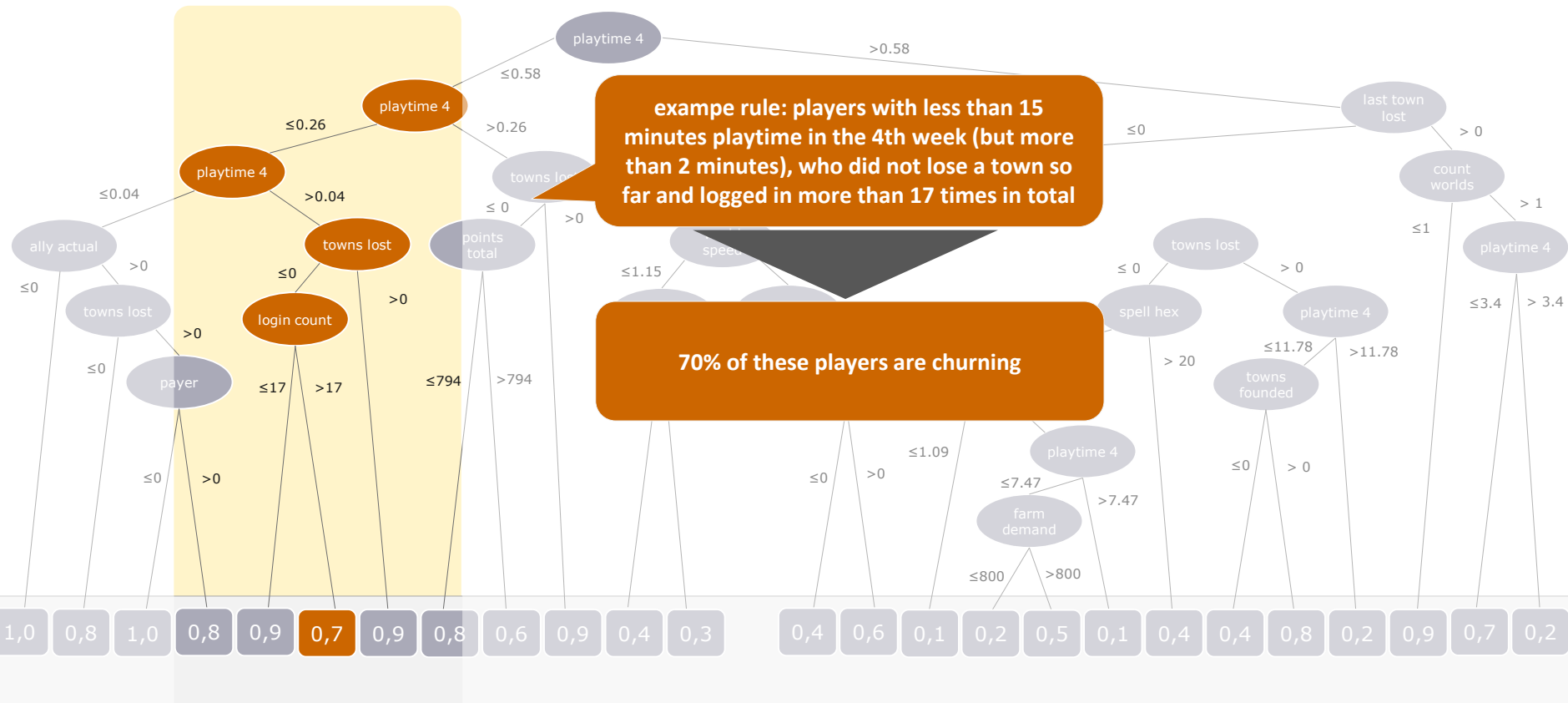
GREPOLIS





actually quite simple: a decision tree model

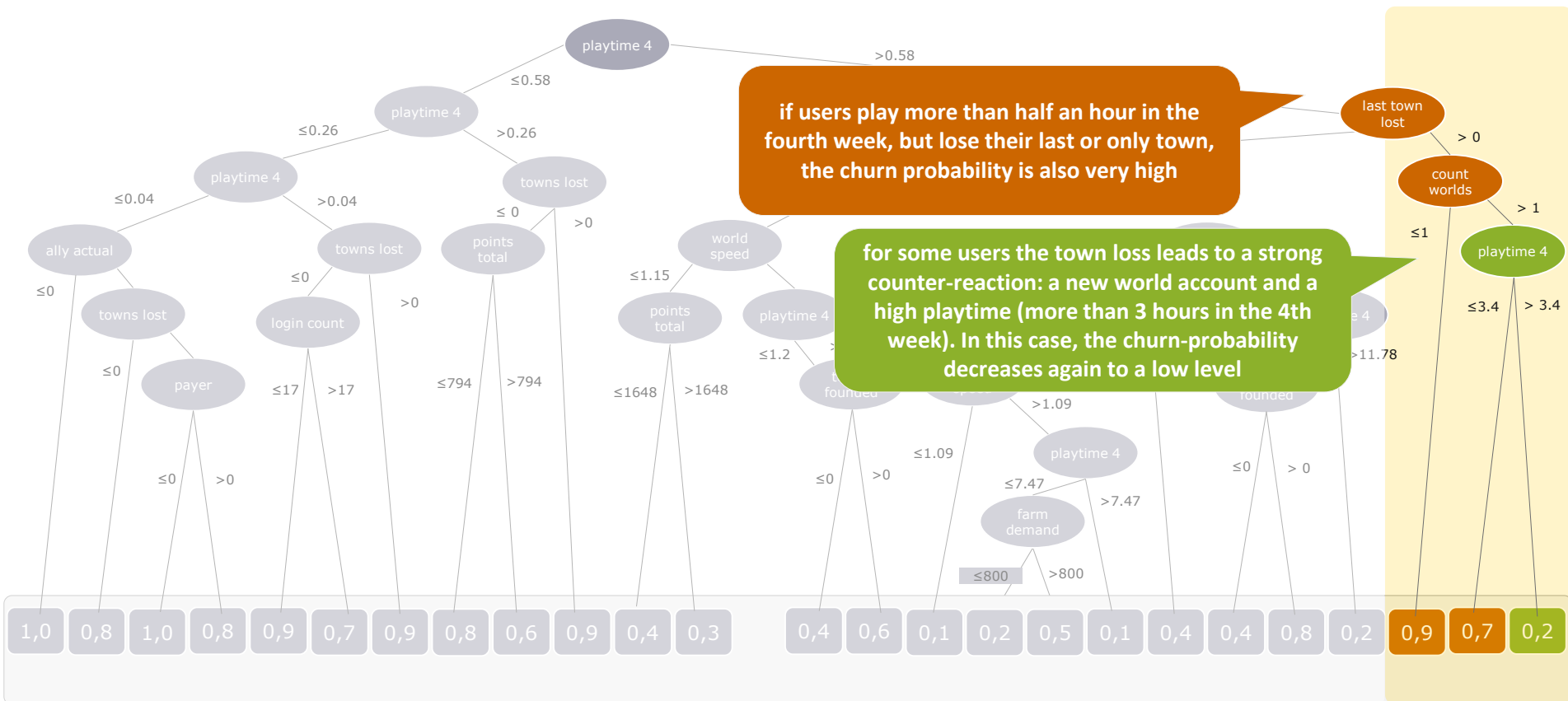
GREPOLIS





actually quite simple: a decision tree model

GREPOLIS





prediction power

GREPOLIS

model for mid-game players
(after 4 weeks of playing)

28% of churners are
wrongly predicted as
non-churners

		reality	
		no churn	churn
prediction	no churn	70%	28%
	churn	30%	72%

30% of non-churners
are wrongly predicted
as churners

model-hitrate
71%

less playing data = less chances for separation
(e.g. newbie-protection: no fights)

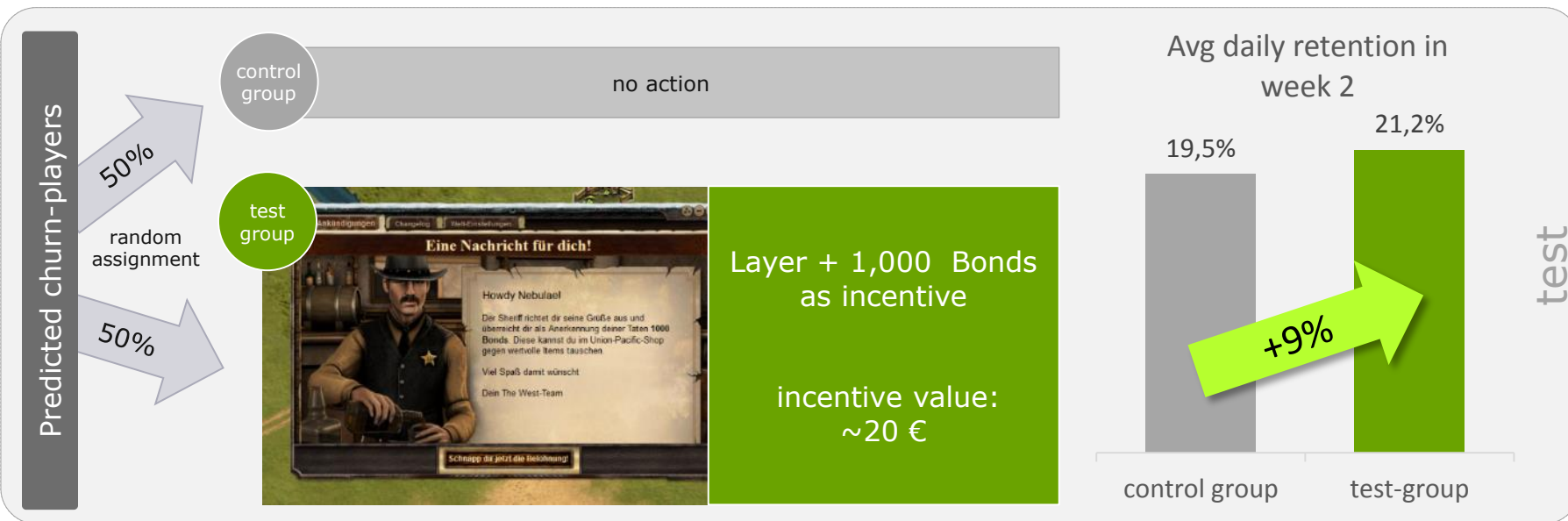
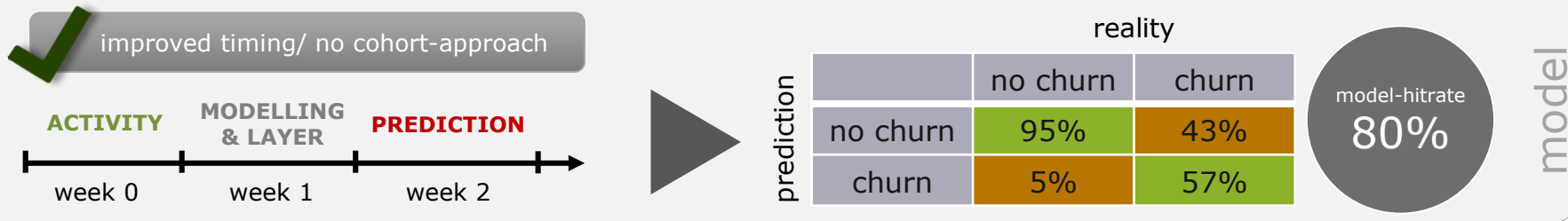
model for early-game players
(after 1 week of playing)

		reality	
		no churn	churn
prediction	no churn	65%	35%
	churn	35%	65%

model-hitrate
65%

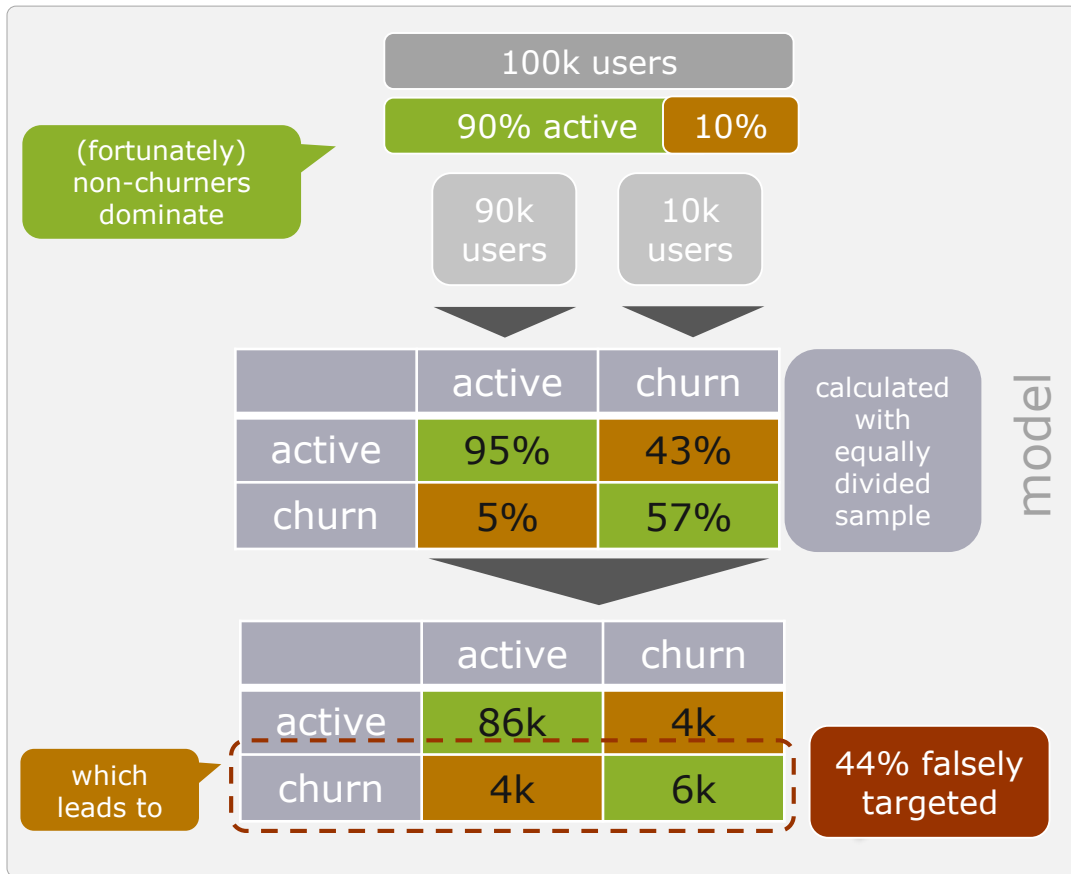


first test: nice start ...





... but a lot potential



THE WEST



Sheriff seems to be rather sceptical that you get a reward ...

style

only 63% of the users who received the incentive cashed the reward afterwards → user-flow (inventory) was unclear for churn-candidates

usability



second test



improved model/ scoring:
more stable model, based on
broader data-set, boosted
towards churn-player detection



improved layer appearance:
more friendly appeal



A/B Test of layer-elements:
with red-eye-catcher vs.
without

	active	churn
active	98%	71%
churn	2%	29%

1st test
80%

model-
hitrate
91%

model

32% wrongly targeted
(42% for first test)

(but a lot of lost potential – see
false positives)

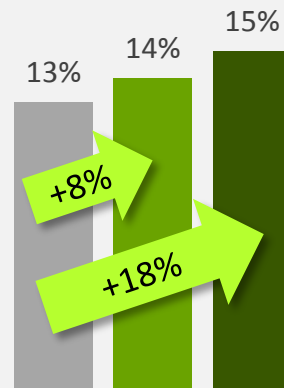
Group
1



Group
2



Avg daily Retention in week 2



■ Control-Group (no action)

■ Test Group 1 (Layer + 1,000 Bonds)

■ Test Group 2 (Layer + eye-catcher + 1,000 Bonds)

test



model optimization

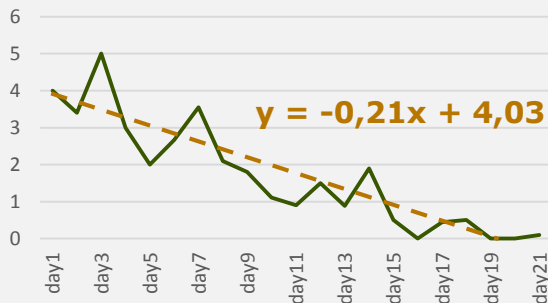
old

Aggregated user-states over a certain-time frame

Churn = no login in a particular time-period

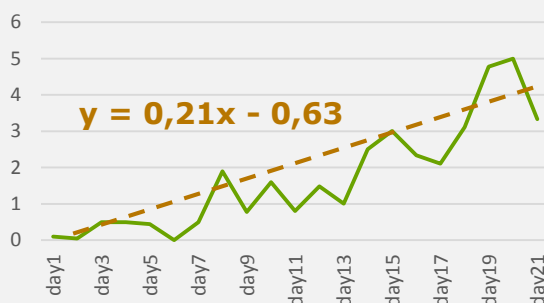
new approach

daily playtime [in hours] of **player A**



playing time last 3 weeks:
35 hours

daily playtime [in hours] of **player B**



playing time last 3 weeks:
36 hours

old approach would consider both users as almost identical; changes in behavior/ playing trend are ignored



Integrate variables for every user which reflect activity development



softer churn-criteria

not "the player has left the game" but
"the player's activity has dropped below the churn threshold"

Threshold: not more than 3 days with game logins during the last 30 calendar days



model optimization

		churn-probability	hitrate	all activity segments are predicted with comparable accuracy
days of play in last 30 calendar days	26-30	→ 0%	100%	
	21-25	→ 11%	97%	
	16-20	→ 32%	93%	
	10-15	→ 47%	92%	
	4-9	→ 66%	91%	
	0-3	→ 91%	94%	



third test



improved model/ scoring:
trend variables, churn-threshold
instead of churn



improved layer apperance



A/B Test of incentive:
premium currency vs. item

	active	churn
active	96%	6%
churn	4%	94%

2nd test
91%

model-
hitrate
95%

model

12% wrongly targeted
(32% for second test)

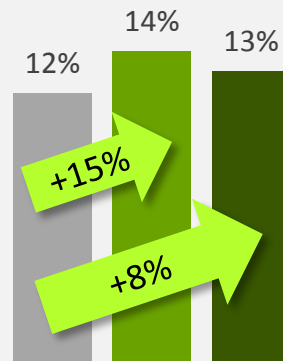
Group
1



Group
2



Avg. daily retention in week 4



- Control-Group (no action)
- Test Group 1 (layer + 1,000 Bonds)
- Test Group 2 (layer + item)

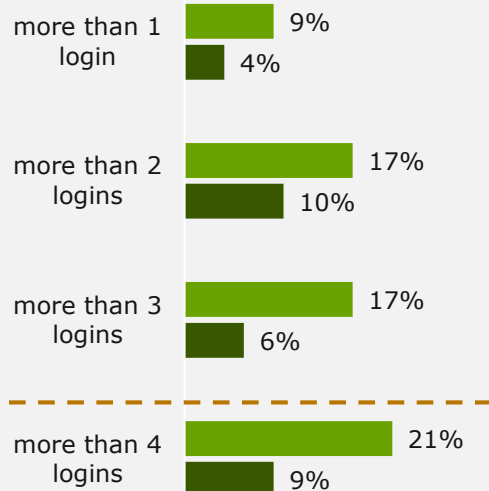
test



more positive details

the stricter the retention-criteria, the higher the impact

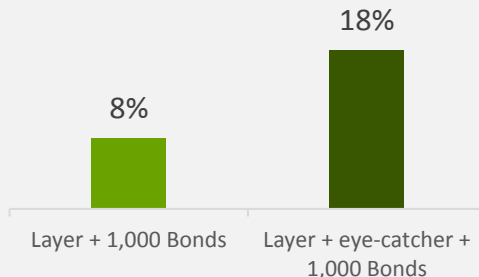
unique retention-plus
(compared to control-group)



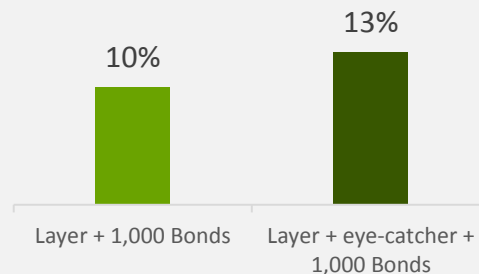
■ layer + bonds ■ layer + items

surprisingly, the impact is sustainable

test 2: effect in **week 2**

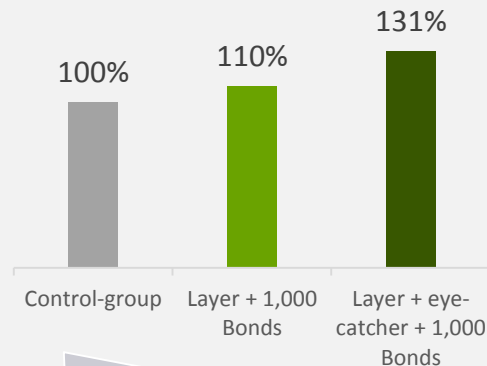


test 2: effect in **week 6**



even though not the target, there is evidence for a monetization-effect

test 2: lifetime-value per test-group



numbers should be taken with a grain of salt as variance for revenue numbers is high.

roll-out



incentive value ~20 EUR



no
significant
effect

incentive value ~2 EUR



+5%
retention

incentive value ~8 EUR



+9%
retention

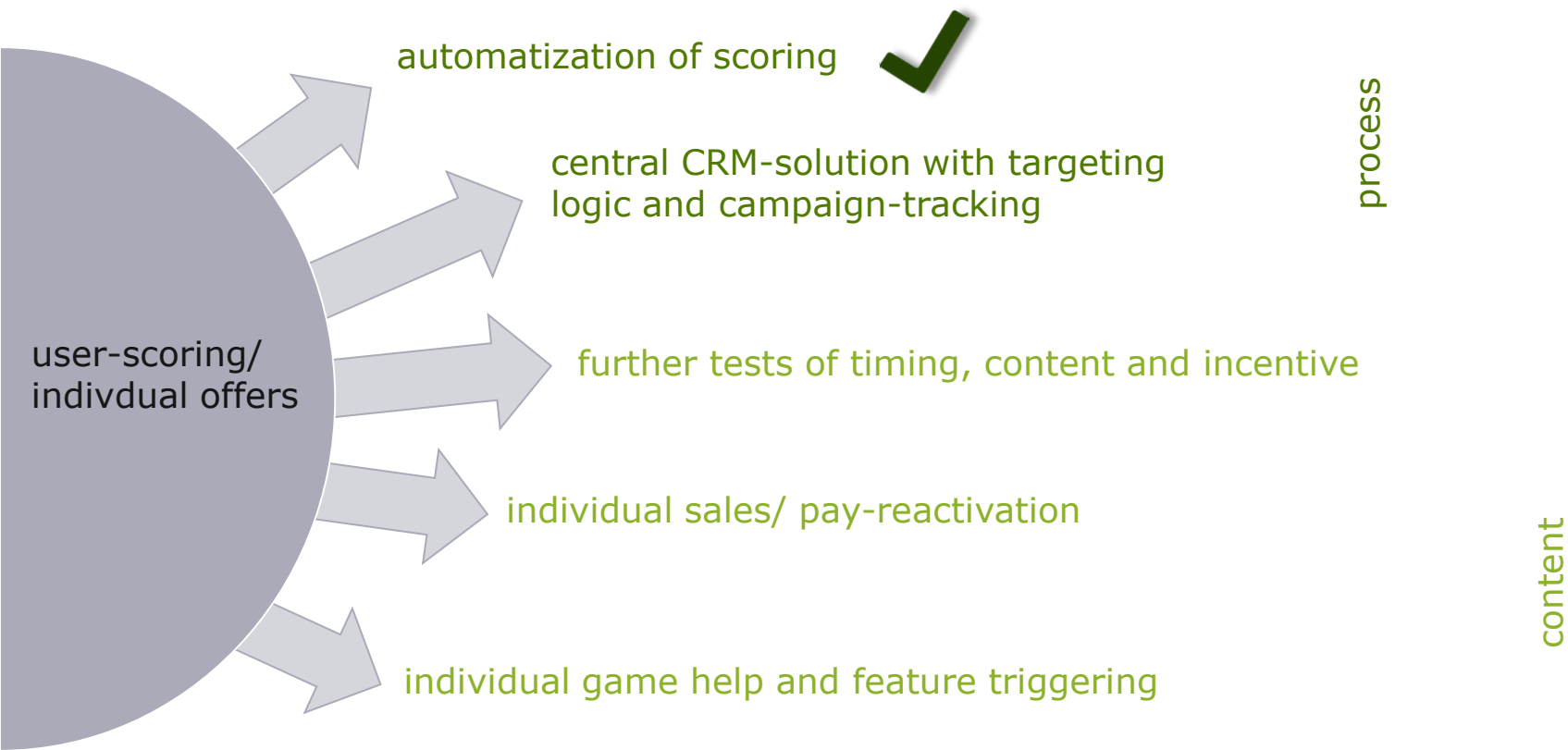


learnings

- » **In general:** Don't wait for a perfect all-around solution, just start testing with the existing data and possibilities
- » **For event-tracking:** Every additional bit of information helps – track the most important events and expand from there, do not start million €-projects that will deliver results in years
- » **Also for event-tracking:** take your time for data-QA – the first implementation will have many flaws. Play the game yourself and monitor the results
- » **For modelling:** Include behavior changes and playing trends as input for your models/ target churn threshold instead of 0 logins. Think about the right timing for the process
- » **For messaging:** Little variations can have a huge effect – test every element of the interstitial
- » **For incentives:** The incentive should give the user enough freedom so that he starts exploring the possibilities of your game



outlook/ next steps

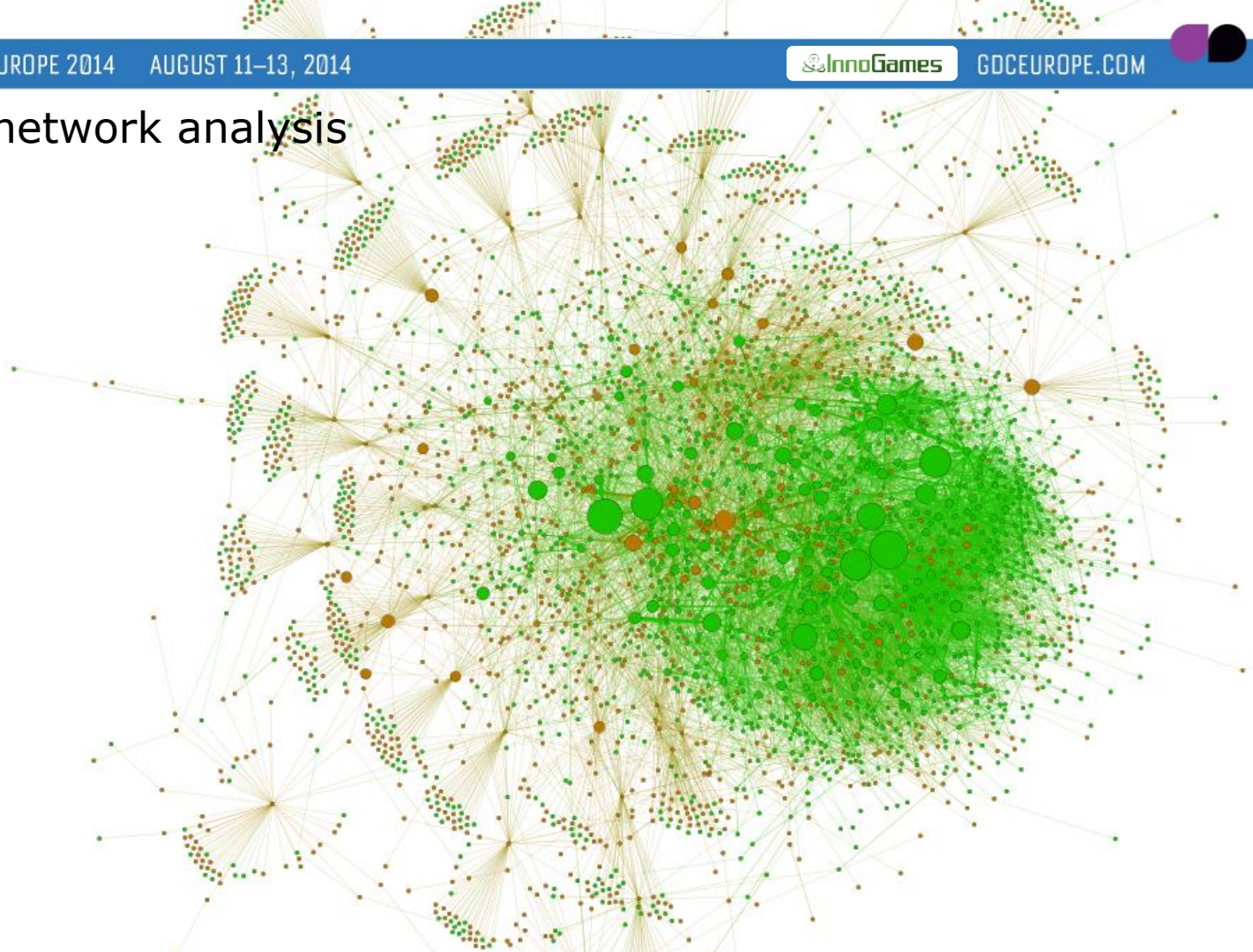




outlook/ social network analysis

Tribal Wars

communication-ties
between users of one
world





thanks to ...

- ... the **analytics team**, especially **Christoph Scholtysik**, for the great work on this and every other project
- ... the **CRM team**, especially **Thomas Cartwright**, for the excellent collaboration
- ... the **game teams** for implementing all this stuff

and thanks for your attention!

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Entertainment booth: Hall 10.1, C15



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