#### DATA INFORMED GAME DESIGN

YASIN HATIBOĞLU

Understanding the player behavior and designing experiences

#### **About me**

- Born in Turkey, studied in Austria, living in Finland
- Education
  - BSc in Econometrics
  - MSc in Quantitative Finance
- Industry experience
  - Data Analyst at Sproing (Austria)
  - Data Analyst at Seriously (Finland)
  - Product Manager at Rovio (Finland)
  - Data Analyst at Supercell (Finland)

#### **Overview**

How can we make better puzzle levels with the help of data?

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How can we make better puzzle levels with the help of data?

- I. How can we identify good levels? (the million \$ question!)
- II. Which mechanics should we have in the levels?
- III. How should we design the level objectives?
- IV. How difficult should the levels be?
- V. How long should the levels be?

## Problem

How can we make better puzzle levels?

#### **User perspective**

Fun levels where players neither fail miserably, nor get stuck or win constantly.

#### **Business perspective**

Core content that generates high resource sink which will lead to IAPs, while not causing high churn.

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Aristotle

"What makes a level good?"

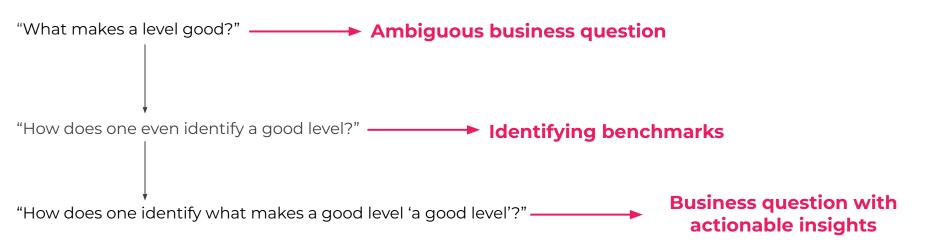
Asking specific, measurable and actionable business questions

"What makes a level good?" — Ambiguous business question

Asking specific, measurable and actionable business questions



Asking specific, measurable and actionable business questions



## Problem

How can we identify good levels?

- We want to identify levels that are performing better and make more of them.
- We also want to identify poorly performing levels and make less of them.

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#### Descartes

"How does one even identify a good level?"

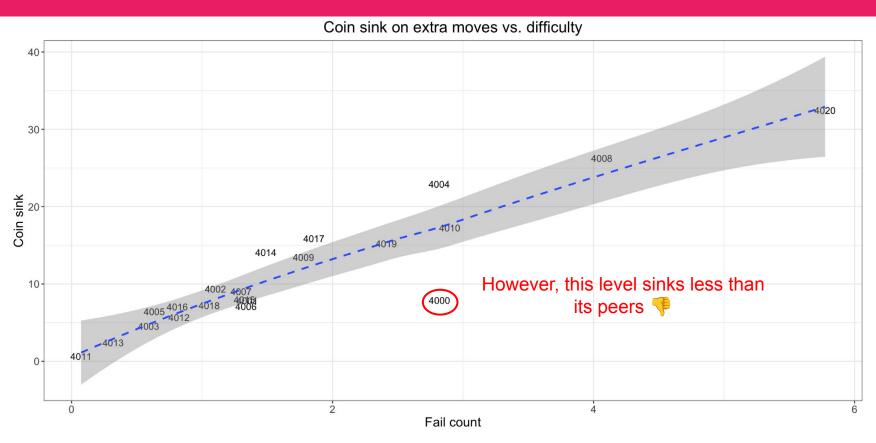
#### **Key level metrics**

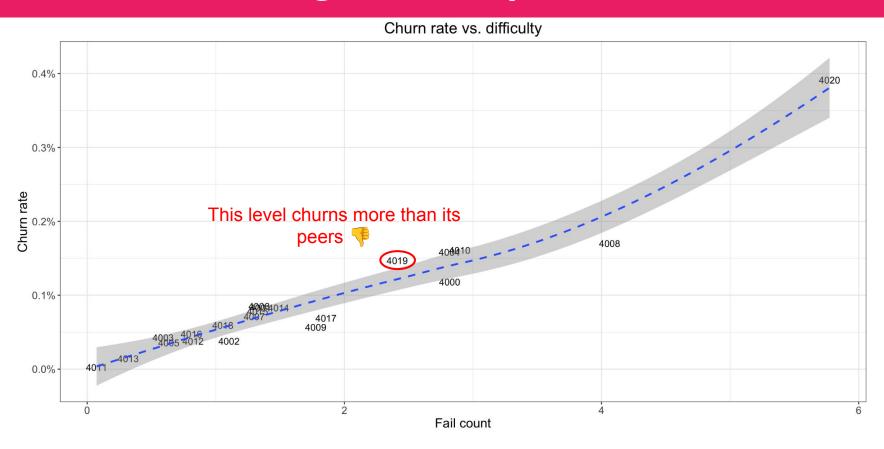
- Extra moves usage
  - Extra moves can be purchased when players run out of moves on a level.
- Difficulty
  - How many attempts does it take to win a level?
- Churn
  - What percentage of players become inactive after starting a level?

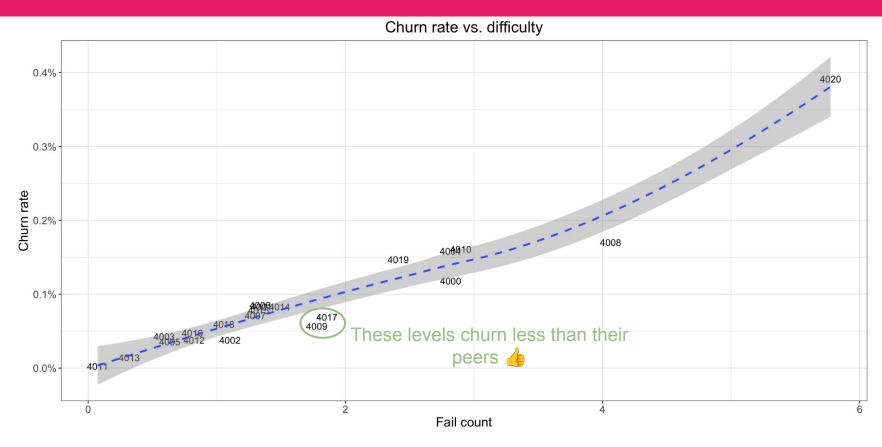












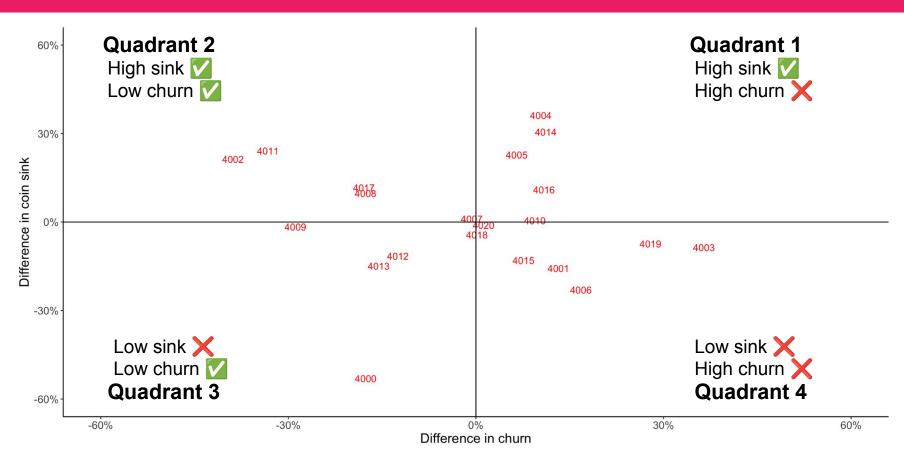
### Act based on the content performance

#### **Content performance matrix**

	Low churn	High churn
Low sink		Worst levels 😿
High sink	Best levels 🤩	

#### # Possible action points

- 1) Identify the "best" and "worst" content/levels. Analyze and look for the common patterns. What could make these levels better or worse than others? Update your design guidelines based on your findings.
- 2) Increase the difficulty of best performing levels. These levels sink more resources while having a lower churn. These levels will bring LTV growth.
- 3) Decrease the difficulty of worst performing levels. These levels potentially have a negative impact on the LTV growth as they churn more than what they sink. It is also recommended to **replace/redesign** this kind of underperforming content with better ones.



#### Possible action points

- Remove the underperforming levels and replace them with levels that resemble the well performing ones.
- Now that we can classify levels into 4 performance quadrants, we can research what level design patterns and mechanics correlate with each quadrant.
- This will help us
  - make more of the good levels (high sink & low churn)
  - o avoid using underperforming mechanics (low sink & high churn).

## Problem

Which mechanics should we have in the levels?

- Now we know which performance quadrant each level belongs to.
- Next, we will use a decision tree to map level mechanics to performance quadrants.

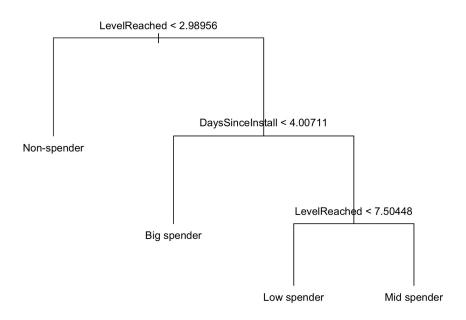
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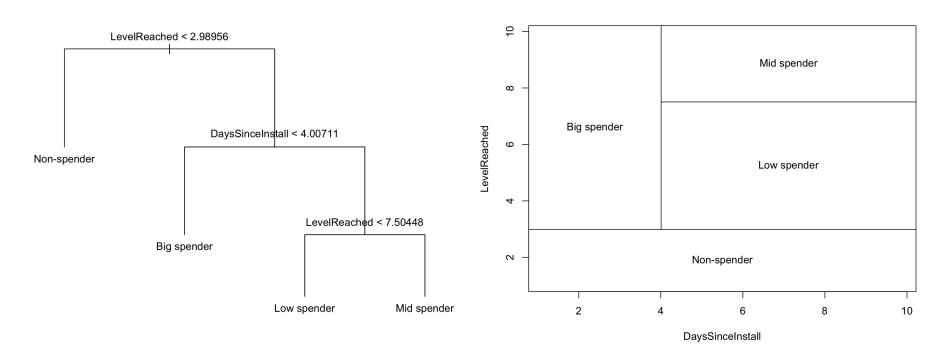
#### Nietzsche

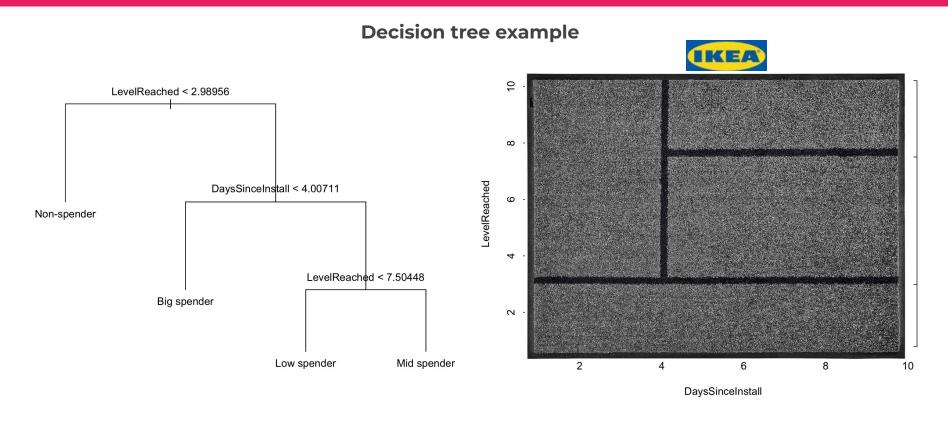
"How does one identify what makes a good level 'a good level'?"

#### **Decision tree example**



#### **Decision tree example**





## Problem

Can we use level design practices to predict the level performance?

**Quadrant 2** 

 **Quadrant 1** 

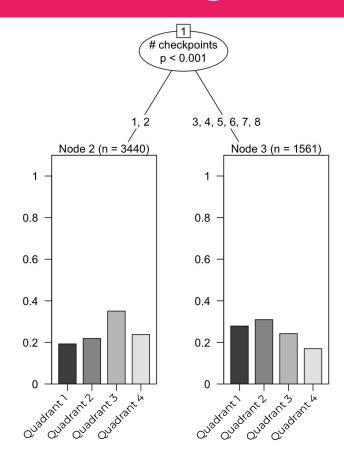
High sink MHigh churn

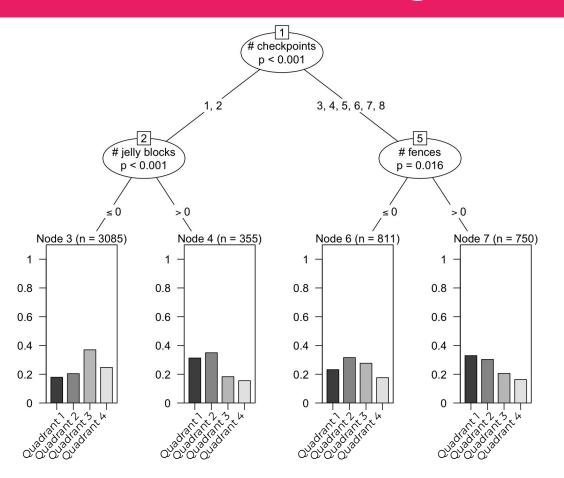
**Quadrant 3** 

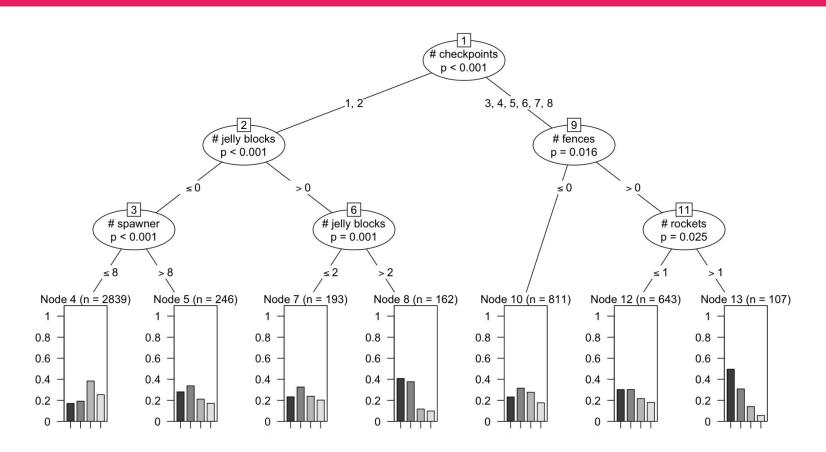
Low sink X Low churn V

**Quadrant 4** 

Low sink X High churn X







# Recap

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Compare resource sink and churn per difficulty bracket; estimate the benchmark and identify over- and underperforming levels accordingly.

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Nietzsche

"How does one identify what makes a good level'?"

Gather data about level mechanics and try mapping them into level performance data. Identify the common patterns on well and poorly performing levels.

Statistical models can be used for initial exploration.

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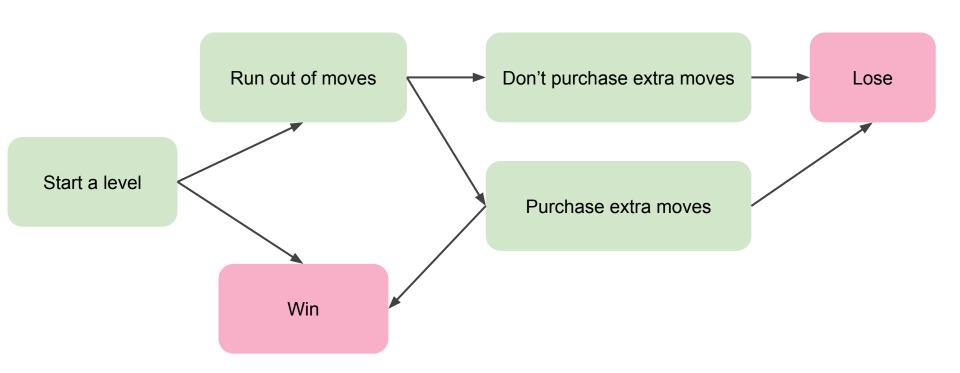


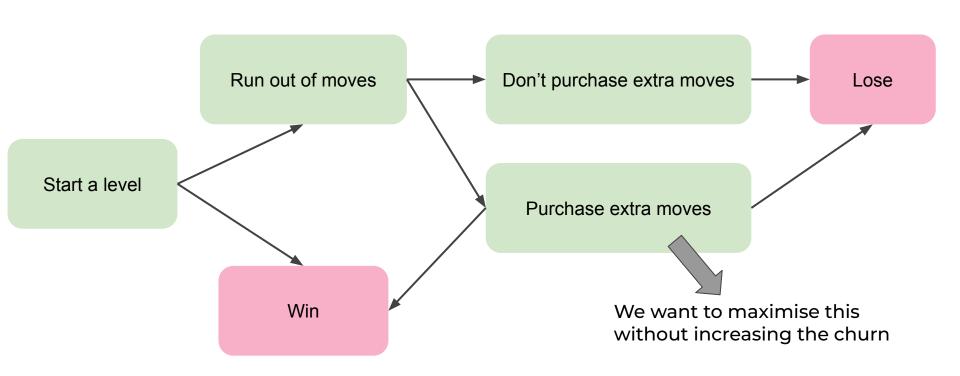
Lil Jon: "What?"

## Problem

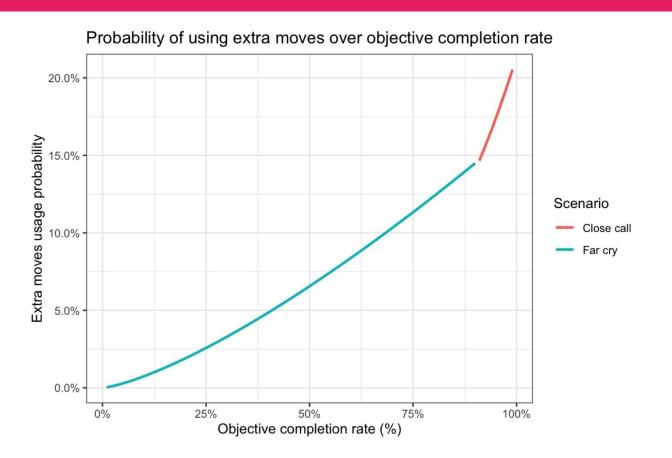
How should we design the level objectives?



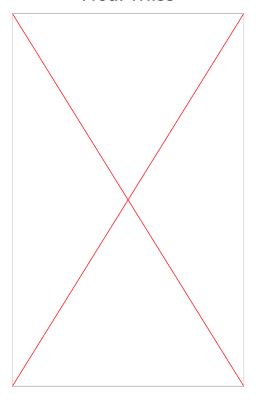




- If we want players to purchase extra moves, we first need them to run out of moves.
- Players' probability of purchasing extra moves is affected by how much players progressed with level objectives.



#### Near miss



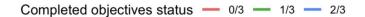
#### Far miss

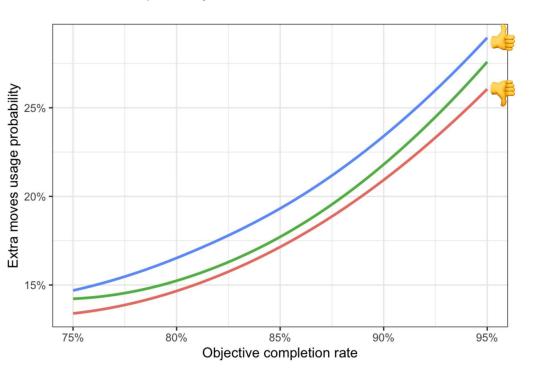




- With a simple UI change, players can get a better feeling of progression.
- Having multiple slugs with varying HPs made the progression less predictable, hence lowered extra moves usage at out of moves screen.
- Replacing multiple slug levels with a single boss slug where the HP was visualized as a percentage scale improved the experience greatly (less churn, higher sink).

Extra moves usage probability over objective completion rate Breakdown by number of completed objectives

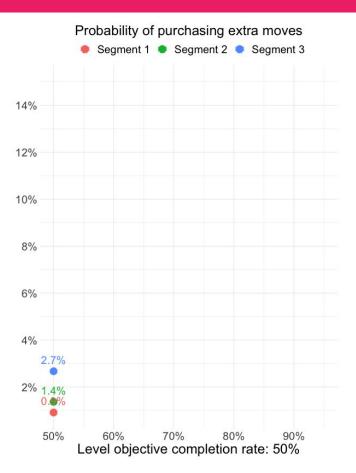




Players don't know what percentage of the objectives they have completed.

However, they can see **how many** of the objectives they completed.

Fewer incomplete objectives gives higher confidence for players that purchasing +5 moves will help them win the level.



- Segment 1 seems to be a very conservative player group, probably non-payer.
- Segment 3 is the least risk averse player group, probably a big spender.

#### **Highlights and action points**

- Levels should be designed in such a way that players are not too far from level completion when running out of moves.
- Having multiple slugs, scenes, checkpoints and multiple objectives emphasize the progression.
- Levels with multiple objectives and scenes usually perform very well...
  - ...**only if** players are at the last screen or last objective when running out of moves!
- Smart algorithms that carry players to higher completion rates can improve both retention and monetisation.

# Problem

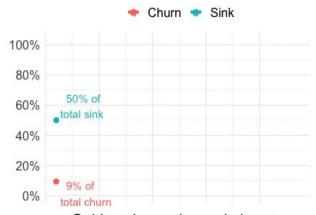
How difficult should the levels be?

- There is a positive correlation between difficulty and resource sink.
- Harder levels are needed every now and then to increase the extra moves usage.
- It is however beneficial to lower the difficulty if players start getting stuck!

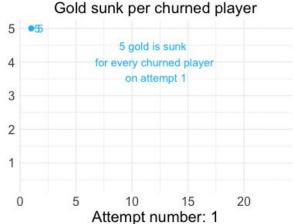
#### We want

- higher number of out of moves screens (extra moves usage will increase)
- fewer cases where players get stuck in a difficult level (churn will decrease).

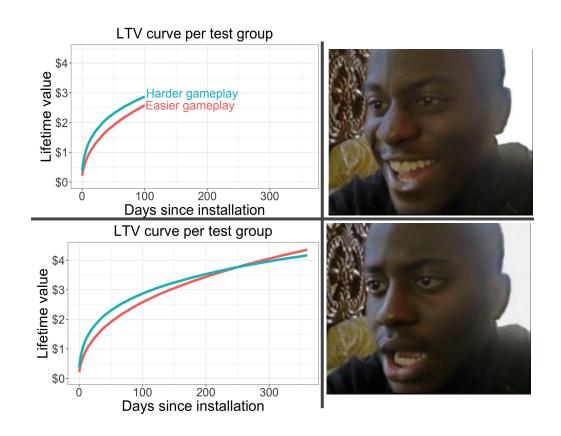
#### Cumulative proportions over attempts



Right after the first attempt, we start seeing diminishing returns in terms of extra moves usage and churn trade-off.



Most of the extra moves usage occurs at the first attempt, while churn starts escalating over attempts.



#### **Highlights and action points**

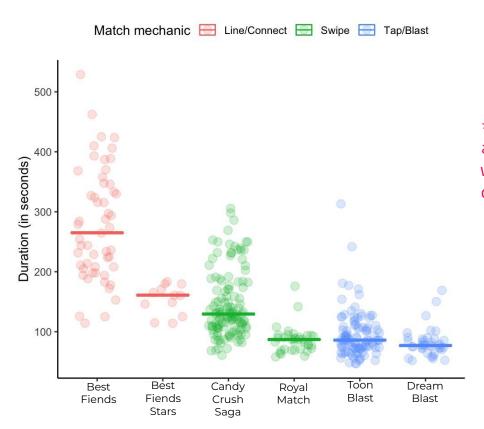
- It's important to have multiple easy levels in between very difficult levels, so that players can build up their win streak. Keeping the win streak going is one of the biggest motivators for purchasing extra moves.
- Variance of difficulty within a level should be low. Higher variance might mean that players either win too quickly or get stuck in the level.
- To avoid players getting stuck, a dynamic difficulty system can help players in a subtle way after a certain amount of failures.
- However, the helping system shouldn't be predictable, since players can exploit it by failing on purpose to receive help.

# Problem

How long should the levels be?

#### **Determining the level durations**

Median level attempt duration per game (in seconds)



\*Level durations are approximated by walkthrough video durations on Youtube.

### **Determining the level durations**

#### Long vs. short level attempts

Long attempts	Short attempts
Higher "sunk cost feeling" drives more extra moves purchases <b>per level.</b>	Lower "sunk cost feeling" drives less extra moves purchases <b>per level.</b>
Fewer levels played <b>a day*.</b>	More levels played <b>a day*.</b>
Fewer out of moves screen impressions.	More out of moves screen impressions.

<sup>\*</sup>daily playtime is not affected by average level durations

#### **Determining the level durations**

#### **Highlights and action points**

- AB testing levels with different durations to find the sweet spot.
  - Figure out if many short levels or a few long levels a day yield higher LTV.
- In Clash of Clans, one could easily sink \$100 worth of hard currency in one session, while in match 3 games, sink opportunities are bounded by level attempts.
  - o In addition to the core level content, a meta layer can increase the possible sink points.

# Thank you!

# Questions?