

Game Biz

Tiago Tex Pine

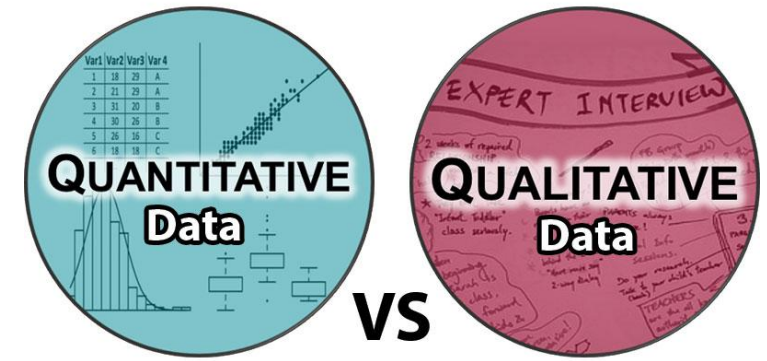
@texpine



Data Analysis for Games

Beyond averages

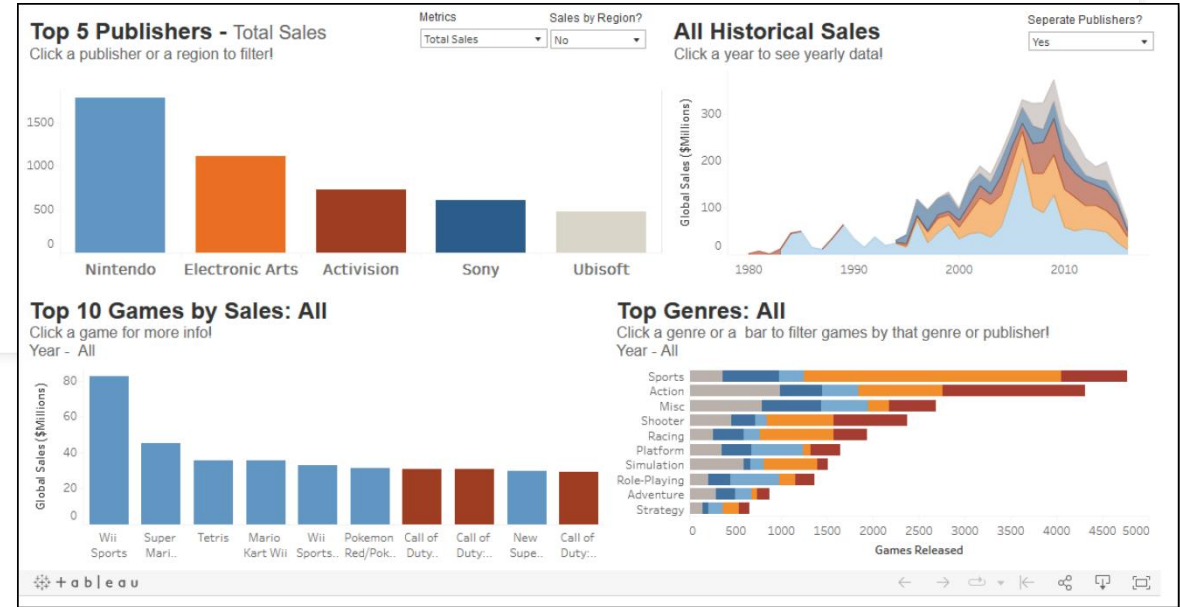
How to use Data



- Two kinds:
 - **Qualitative**: opinions. User surveys, playtests, usability labs, Twitch, experience. Cross multiple dimensions of pattern in the mind and tells you **why things happen**.
 - **Quantitative**: real user behavior and actions in the game. Data tracking and analytics. Tells you **what is happening and how**.
- **Both** are necessary.
 - **Qualitative** data is the classical way, player impressions.
 - But **Quantitative** is the real behavior not distorted by perceptions.
 - *Example*: when some complain about metagame imbalance, they could be just angry. Data can resolve if it's real in the universe of all matches.

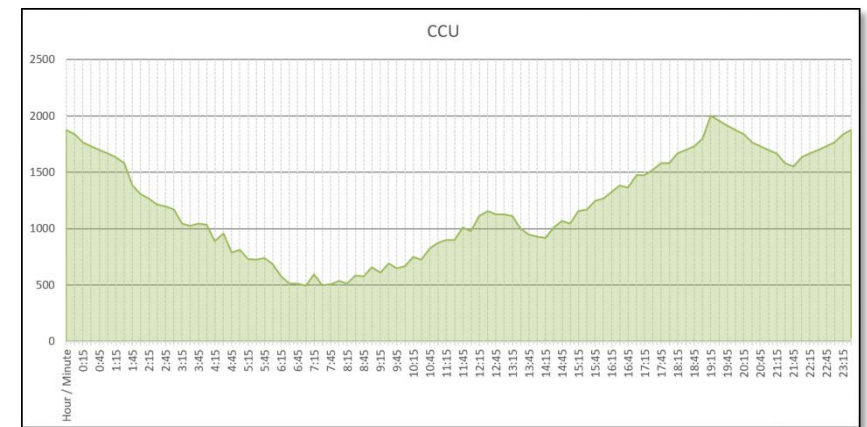
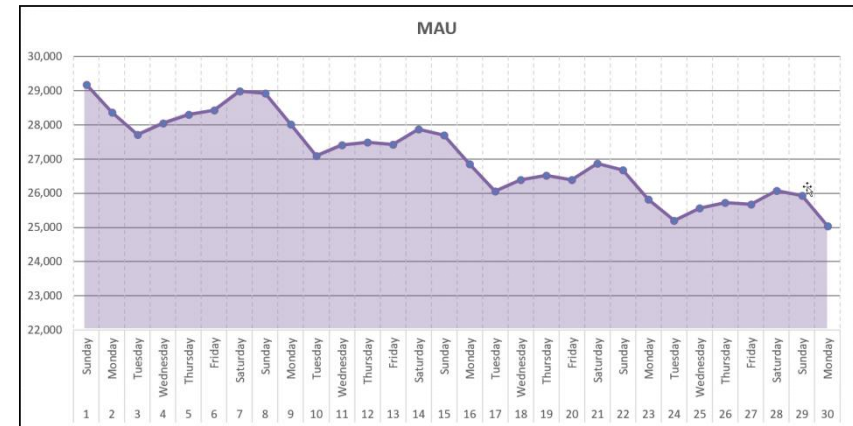
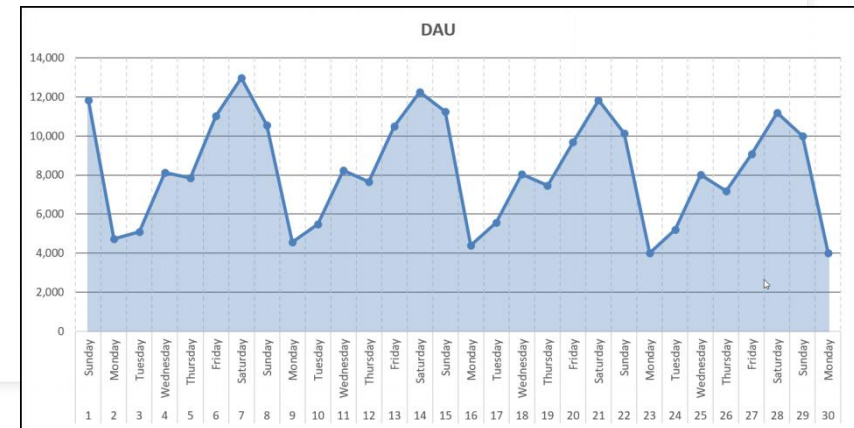
How to use Data

- Pure data is just trivia: useless.
 - It needs to be compounded with actions.
 - "*Data is the new oil*" - but oil is only useful after processing, refinement and manufacturing.
- Needs to be followed by actions.
 - Analytics cannot *decide* directions for the game, they can only inform.
 - Analytics cannot *design* features or loops, they can only measure those that exist.
- This means any investment in data tracking, analysis or science needs to be integrated into decision-making systems, otherwise it's a waste of resources.
- Let's see some key KPIs - Key Performance Indicators for games.



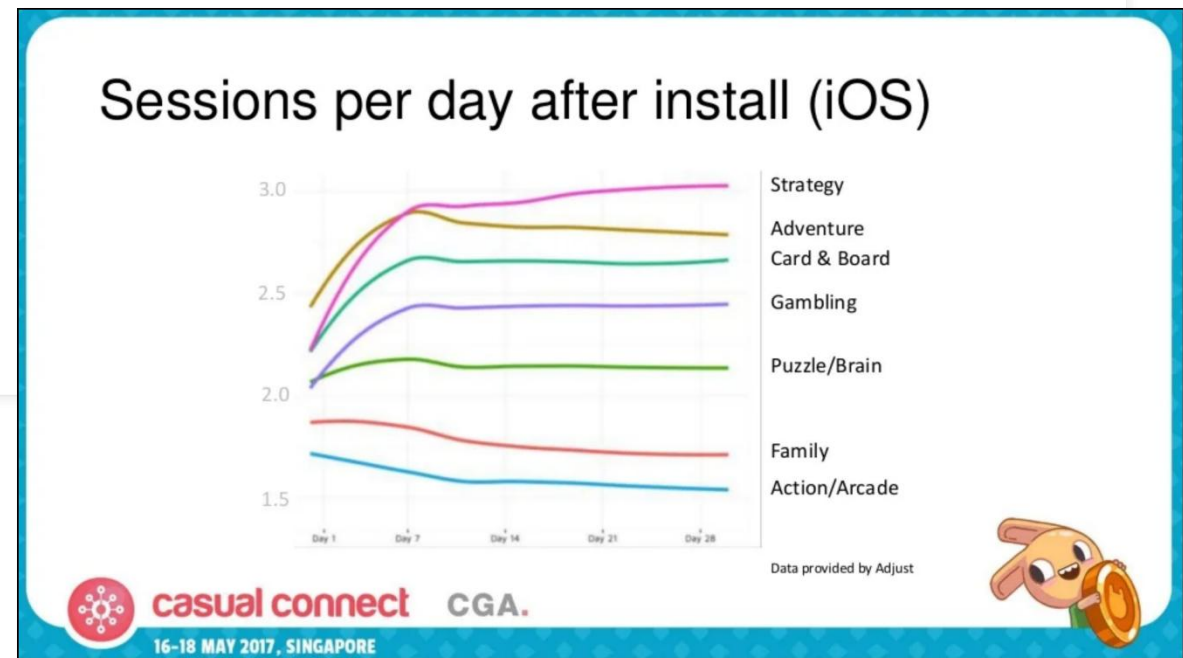
Main Usage KPIs

- **DAU** - Daily Active Users.
 - Count of unique user IDs in a whole day.
 - Daily health of activity of the game as a business
 - Fluctuates down quite a bit on Mondays and up on holidays.
- **MAU** - Monthly Active Users.
 - Count of unique user IDs in a 30-day rolling window.
 - More stable measure of the game as a business.
 - If it's going down for several weeks, a severe business trend.
- **CCU** - Concurrent Users.
 - Measured in intervals of 5 to 15 minutes across the day.
 - Useful for backend stability. **Peaks** matter more than **Averages**.
 - Also informs Live Ops on intra-day behavior, which is useful for notifications and short Events.



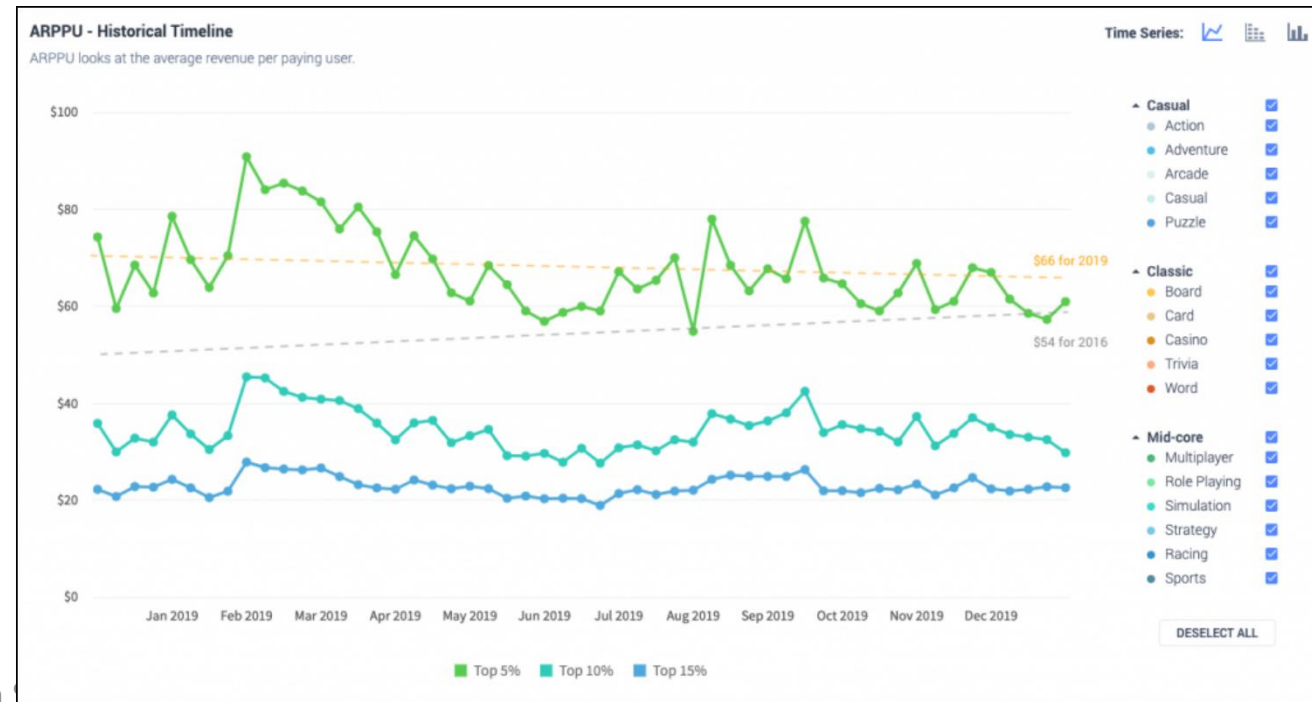
Engagement KPIs

- **DAU / MAU Ratio = stickiness.**
 - Good stickiness start at around 0.2 (20%).
 - Games should not be lower than 0.15. Lower indicates retention issues.
 - Top of the line: Facebook and social media reaches 0.5 (50%).
- **Sessions per Day / per Week**
 - Average amount of sessions per user, per day of activity. For mobile games, normally between 3 and 5.
 - Could also be measure per week for a more stable behavior, but less useful for heavy Live Ops.
 - Useful for design of Live Ops features as well.
- **Session Duration**
 - Average minutes of gameplay in a session. For mobile games, anywhere between 7 and 25 minutes.
 - Very good information for game designers, particularly for stage/core loop duration and Appointment mechanics.
- Beware those can **change significantly in different points of the player progression.**
- **Installs / New Users**
 - In premium, same as purchases. In free-to-play, very different. For both, key for marketing investments.



Main Revenue KPIs

- **ARPU** - Average Revenue per User.
 - Measured as a ratio of Revenue / Amount of users ratio.
 - Could be calculated on different time windows: monthly, daily, yearly or lifetime (all users since the App launched).
 - When Daily Revenue / DAU, also called **ARPDau**. For free-to-play mobile, between anywhere \$0.05 and \$0.50.
- **ARPPU** - Average Revenue per Paying User.
 - Measures how much a paying user is willing to pay.
 - For Premium games, buyers + DLC.
 - For free-to-play games, ideally US\$ 30 or more.
 - Top games have this US\$ 70+, all the way up to \$500+.
- **Conversion Rate**: % of users who purchase something.
 - Users who purchases something in the game at least once.
 - Calculated on different time windows.
 - *Daily Conversion* is good to measure short-term impact of promos.
 - We want this above 1% for *lifetime* conversion.
 - Top games can reach 5%+.



Retention

- Measure of how many users come back at N-days-after-install.
- There are a number of ways to calculate Retention. The most important ones:

CLASSIC

MON	TUE	WED	THU	FRI	SAT	SUN
10	3	2				
8	9	10	11	12	13	14
15	16	17	18	19	20	21
22	23	24	25	26	27	28
29	30					

+ INDUSTRY STANDARD.

+ EASY TO EXPLAIN.

- SENSITIVE TO DAILY FLUCTUATIONS.

ROLLING

MON	TUE	WED	THU	FRI	SAT	SUN
10						1
8	9	10	11	12	13	14
		1				
15	16	17	18	19	20	21
1	1					
22	23	24	25	26	27	28
29	30					

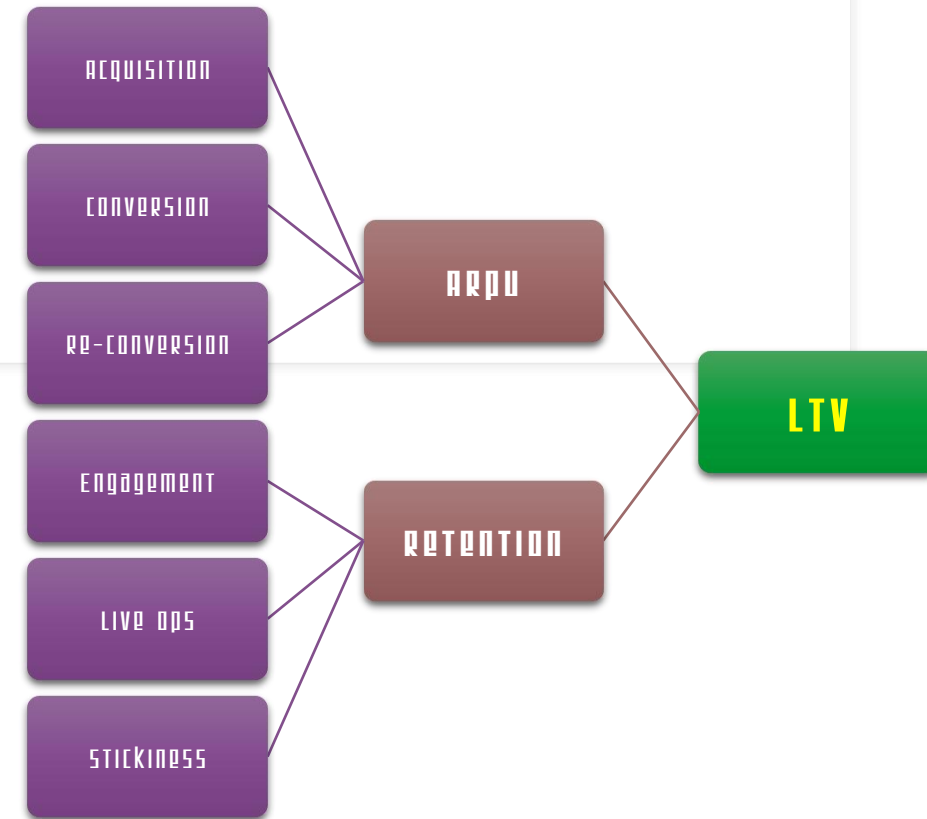
+ FASTER TO CALCULATE: DATE OF FIRST USE AND LAST USE

+ REFLECTS LONG-TERM STICKINESS BETTER.

- POOR FOR DAILY ENGAGEMENT, TREATS A DAILY USER THE SAME AS A USER WHO COMES BACK ONLY EVENTUALLY.

Lifetime Value (LTV)

- A composition of Retention and Revenue to try to project the value of each *future* user.
- **Key** for marketing efforts: you *cannot* spend more on acquiring 1 user than your LTV.
- There are *many* ways to calculate this.
 - The simpler, just a past ARPU, but that's the *past*, as the game was before.
 - A more robust way considers **Churn / Retention** as updates and Live Ops kick in.
 - The time period of all KPIs should be the same: monthly, daily, weekly or yearly, depending on how marketing budgets are assigned.
 - The Churn part is the trickier part of the formula.



$$LTV = ARPU \times \frac{1}{\text{churn rate}}$$

EXAMPLE

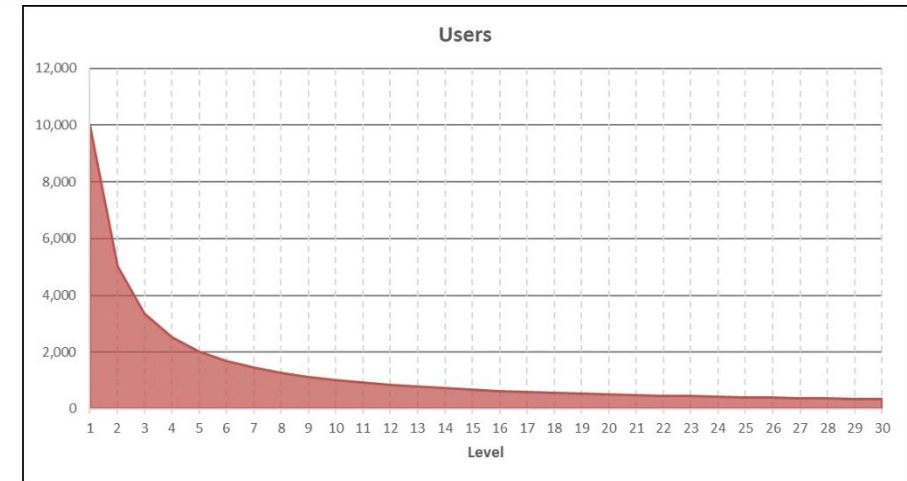
$$LTV = \$1.35 \times \frac{1}{0.6} = \$2.25$$

with a *monthly* ARPU \$1.35 and a *monthly* churn rate of 60%, your average customer moving forward has a predicted LTV of \$2.25.

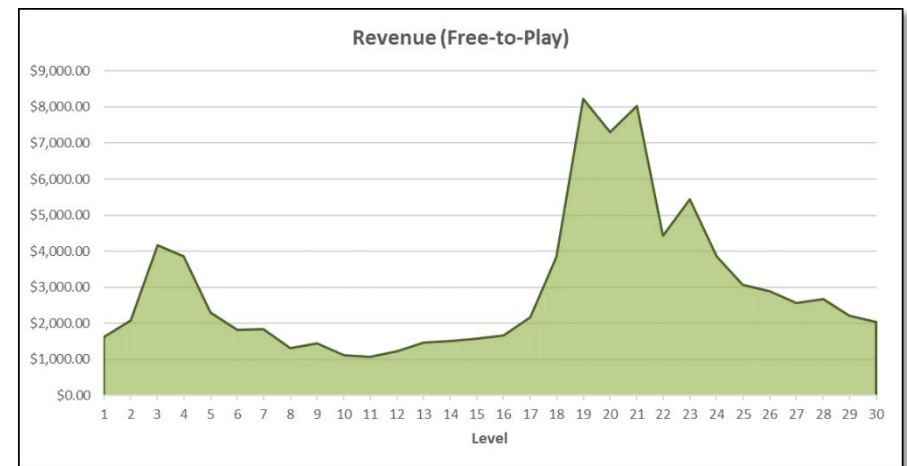
Averages vs. Distributions

- ***Be careful with averages.***
 - Guiding oneself too much with them lose the finer details.
 - Amongst your entire population of players, your Fans are outliers.
- Example: progression and ARPU.
- High-level averages are often demanded by managers,
- ... but what *game designers* need to improve the game are those distributions.
 - They are the one that can locate *where* problems are.

AVERAGE
LEVEL 7.5



ARPU \$2.17





Final words!