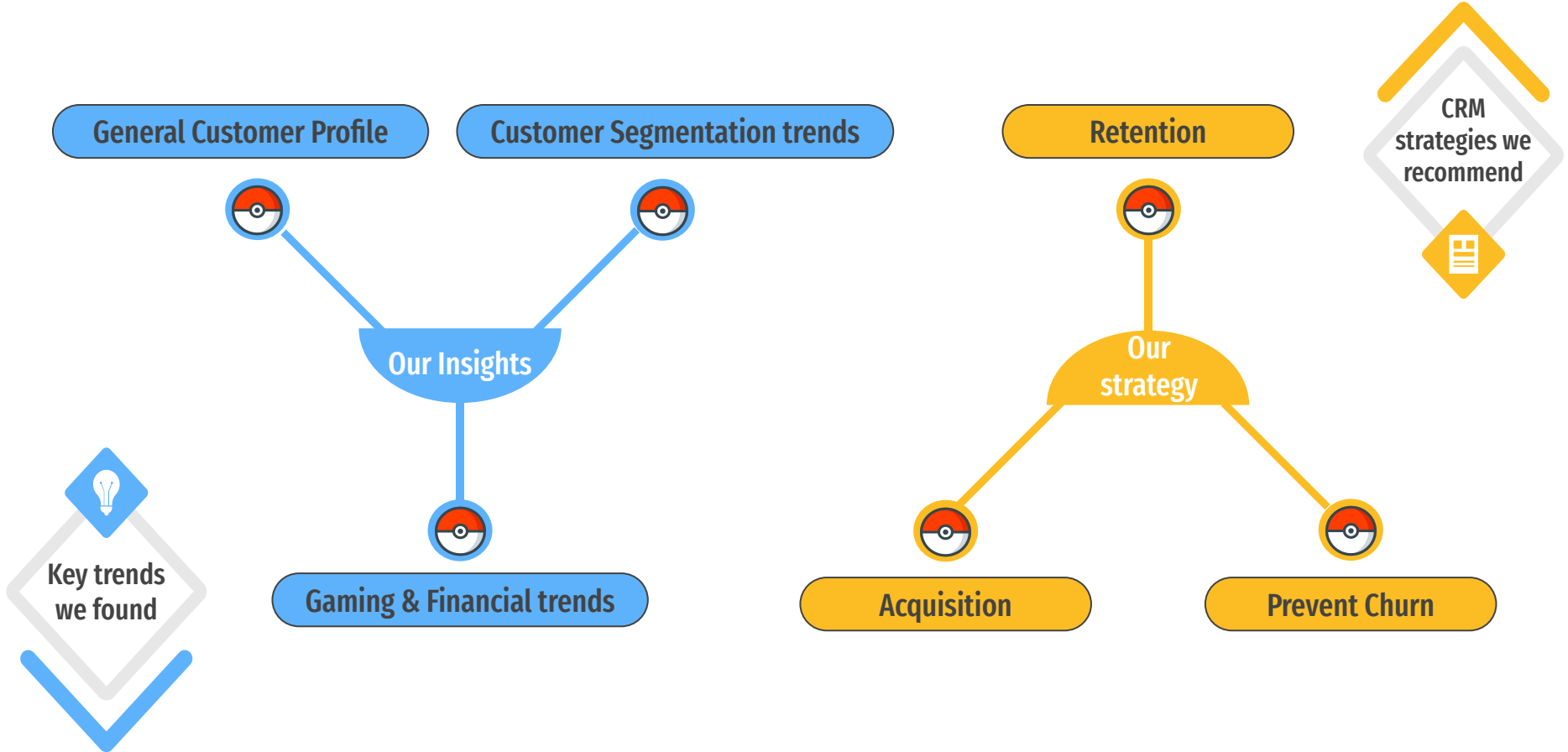


# Pokémon-Go Presentation



Delivering recommendations to Niantic

# In Brief: Delivering value to Niantic



# General customer profile



*All facts are based on summer 2018 player data*

**4,703** are  
active players



**60%** of players  
are male  
players



Median age is  
**27 years old**  
*(75% are less than 34)*



Players with all  
level of income  
are **equally  
represented**  
*(roughly 30 of each)*



**Half the  
players** joined  
the game less  
than a year ago  
*(end of summer 2017)*



# Gaming and Financial Trends



## \$ Financial

Only  $\frac{1}{3}$  of the players bought something in-game

Those purchasers spent **11 euros on average**, majority were males (61%)

Purchasers were not frequent buyers : **95% of them made at most 2 transactions**, 5% made there last transaction during the last summer week



## Both

From our 4,700+ summer active players, **87% did not purchase** anything in fall. (churn rate=87%)

## Gameplay Gaming

**25%** of the players played more than **6** times

On average, their **last playing session** was **a month ago** (assuming we are the 31/08/2018)

In a single gaming session a player walks **4 kms**, has at least **1 social interaction**, plays up to **1 hour** and catches **19 Pokémons**

On average, a player have **5 social interactions** in total during the whole summer

Only **17%** of the churners **received the fall bonus** at the end of the summer.

# Customer segmentation trends



## Age segmentation

Young Adults are the top active and frequent players

Children and Teen-agers are also parts of the top active and frequent players.  
Middle-Age customers represent a significant part of the new players and shall not be forgotten.

*Insights based on an classical age segmentation:  
(Children/Teen-agers/Young Adult/Middle Age Adult/Senior).*

Customers who used to play a lot but are no longer active have the high income level.  
By the end of the summer 2018, the most frequent and active players have the medium income level.

*Insights based on a segmentation by income level of the active players (Low/Medium/High).*

## Income segmentation

More than 50% of new acquired players have low income

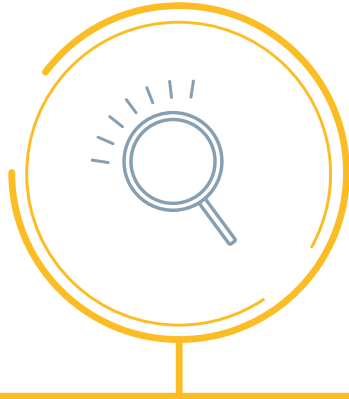
## Customer Type segmentation

Top active and frequent players use Pokemon GO to exercise and have social interactions

Social raiders and Walkers are the most interesting player segments : they play frequently and actively.  
However, Miscellaneous players represent an important part of the new ones the application tends to be no longer used for its features.

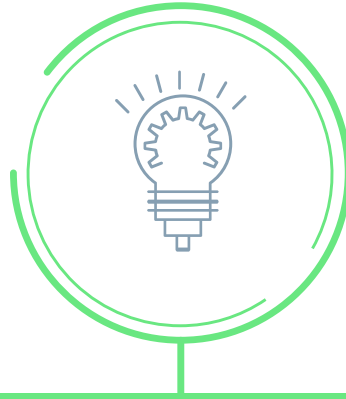
*Insights based on the customer type segmentation made by Niantic (Catchers/Social raiders/Walkers/Miscellaneous).*

## II- CRM Recommendations



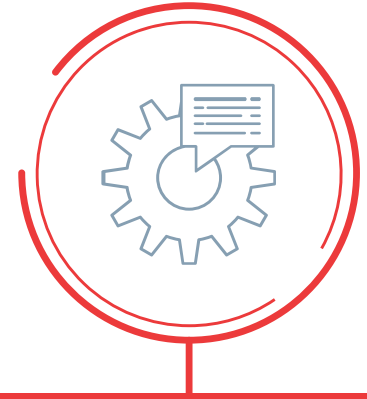
**Acquisition Strategy**

**How to acquire new players?**



**Retention & Development Strategy**

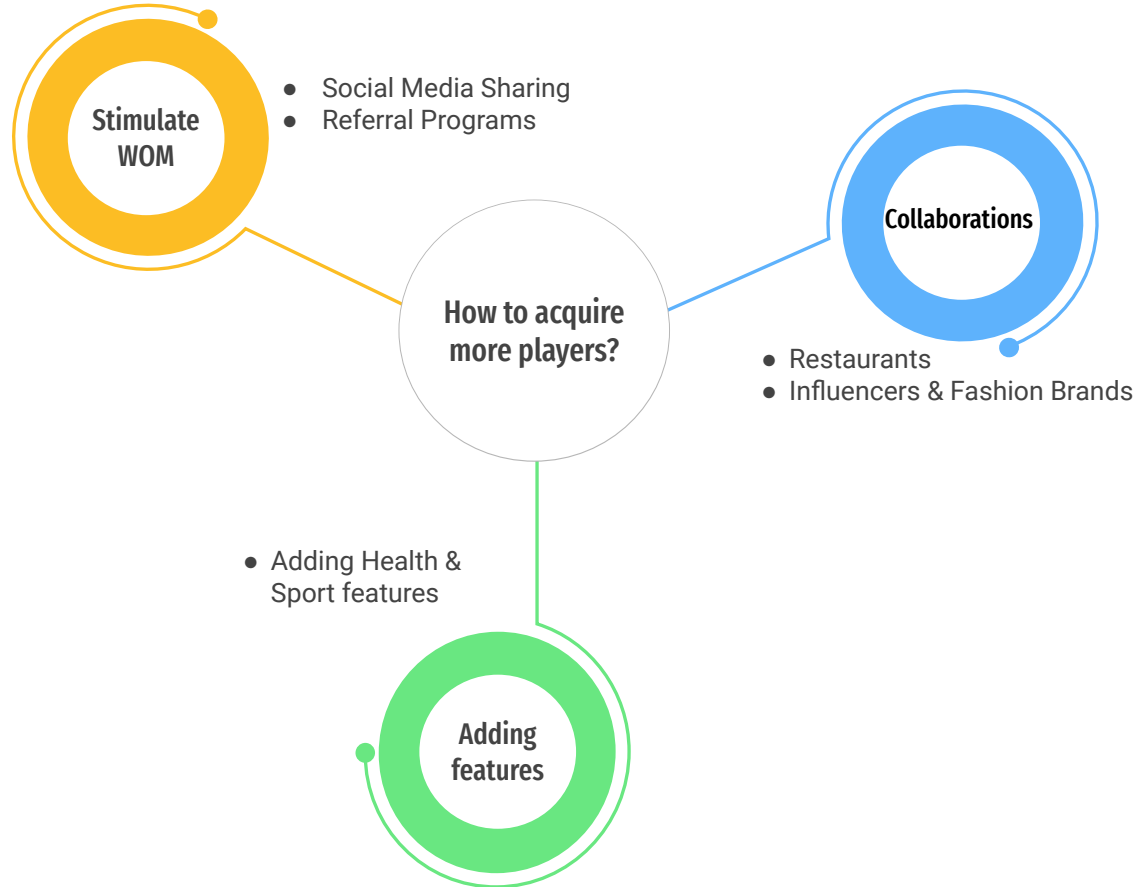
**How to keep the players  
and ensure profitability?**



**Preventing Churn Strategy**

**How to deal with customer churn  
and minimize costs associated?**

# Acquisition strategy



# Retention & Development strategy



## LP based on Transactions :

### Creating 3 levels:

Premium players (1+ transaction)

Medium players (1 transaction)

Basic players (0 transaction)

### Rewards and advantages system:

Getting better rewards according to the levels (based on the LCG financial)



## Strategy of Gamification :

- Separation of the playing map into different territories.
- The best player in each zone wins the territory and gives it his name.
- The "territory" is put back into play each month.

## LOYALTY PROGRAMS



## Emailing campaigns

Adapted to the 3 different types of players **premium, medium and basic players** (defined in the loyalty program)

## TAILOR-MADE TOUCHPOINTS



## Marketing actions

Advertising, direct marketing, promotions, etc.. can give importance in brand equity for retention.

## MARKETING ACTIONS



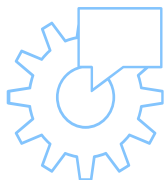
# Preventing churn strategy: predictive model deployment



Our model was trained on 3,300 Pokemon GO players. It is based on their gaming and purchasing behaviors.



**A customized model**



**Probability prediction & classification**



It can predict the probability that an active player will stop purchasing in-game i.e. churning - using different characteristics of the customer.

- Playing frequency
- Amount of money spent
- Customer segment
- Promo received or not



**A performant model**



Model was tested on 1,410 players we knew their churn outcome.

**The model is accurate** : 89% correctly classified players as being churners or not.  
**The model is precise**: 87% of the churn predictions delivered were right.

For a churn prevention campaign of 1 €/customer, with churning players costing 3.4 €/customer to Niantic, we recommend applying a probability threshold of 0.41 in the model.

It will therefore minimize the total costs for prediction failures and increase profitability.



**Improving profitability**





**Thank you for listening !  
Any questions ?**



**Presented by Team 15:**

Charles Lietar

Omar El Hajjar

Charlotte Naepels

Aly Chour

Pauline Chomel de Jarnieu