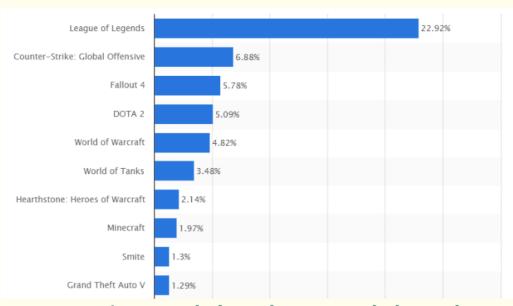
Linhai Zhang Wen Huang Jiacheng Xu **Zong Guo** Yilun Chen Conclusion League of legend Thank You Linhai Zhang Wen Huang Jiacheng Xu **Zong Guo** Yilun Chen Data Analysis Conclusion So for, we get some models with prediction anternacity 60%. Some important variable are selected to help player have better conferrant for the game Variable selection and model parameter adjustment are our next consens. League of legend Thank You

League of Legend



League of Legends has dominated the online market since 2012

League of Legends is big, so big that its player base amounts to more than one percent of the global population.

September 15 2016

Our goal is:

- 1. Finding useful strategy to help players to win easily.
- 2. Giving player some suggestions for different areas and levels.

Game Introduction



What can you do to win a game?

- 1. Player's level
- 2. Players' cooperation & performance
- 3. Composition in each team

Data Discription

Crawling data from LOL official developer

Website: developer.riotgames.com

Patch: 7.6 2017/3/22 to 2017/4/5

Area: South Korea & North America

Database Selection

Pick 1,000,000 sample from South Korea %>%

Choose 100,000 sample from formal rank %>%

Select 10,000 sample from high level players



Variable Discription

Each sample contain 10 (2*5) information (10 player) Each information include 29 variables

Damage -

Minion Kill .

Assist -

Death -

Gold -

Total damage includes Hero, Minion, and Structure

Get gold from minion kill and push the lane

Your teammate kill enemy hero under your assist

You can't obtain the resource from the lane

Use gold to buy equipment and improve yourself

Data Analysis

- 1. Data pre processing
- 2. Model based on 5 factors
- 3. Model discription
- 4. Result

Data pre processing
Scale some variables by game length
Detect missing value
Adjust some variables
(There are 134 champions but more than
400 champion ID)

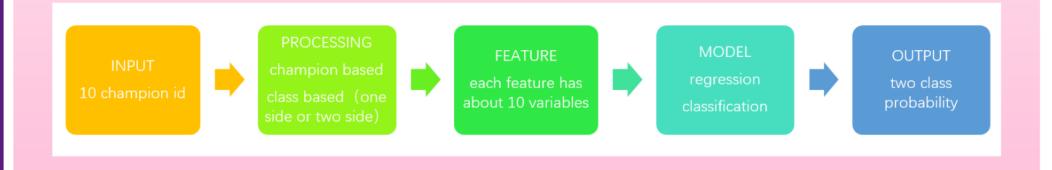
Model based on 5 factors

The first half of the game: stay in position, accumulate resources, not fail in the local war. The rest of the game: team fight!

Five factors:

Top Middle Bottom Jungle TeamFight

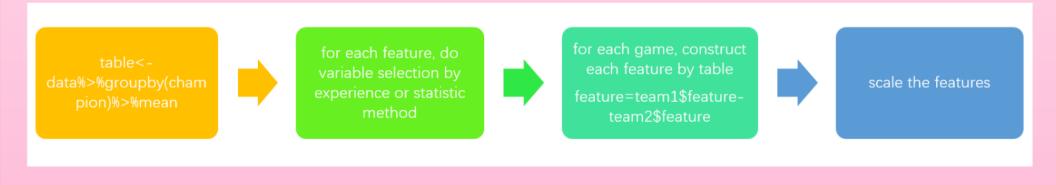
Model discription: basic idea



Model discription: build feature

Model discription: build feature

For champion based



Result: Prediction rate

Method -	Crossvalidation -	Boosting .	All variables	Champion -	Class one side	Class two side
GLM .	TRUE -	FALSE .	51.1% -	52.2% -	52.8% -	58.8% .
Randomforest	FALSE -	TRUE .	51.9% -	51.9% .	52.3%	56.4%
C5 .	FALSE -	TRUE -	51.4%	51.6%	52.0%	56.0%
RIPPER -	FALSE -	TRUE .	50.2% -	51.5% .	52.2% -	55.5% .
Neural Network	TRUE -	FALSE .	50.4%	52.0%	52.5%	53.2%
XGBoost -	TRUE -	TRUE -	51.2% -	53.3% .	53.1% -	59.3% -

PS: The result in most of the papers are about 52~57% and the best one can achieve 67% using the information after the 1st 10 minutes of the game, but what's the meaning to predict the result when the game have already started for 10 minute?

Result: Significant variables for GLM

Top: DamageToken & Damge

Middle: Damage & HeroDamage

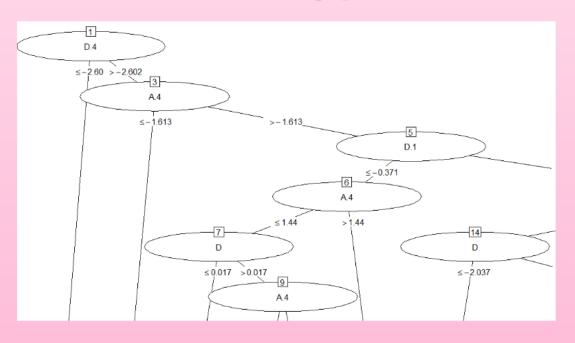
Jungle: Assist & Damage

ADC: Damage & Death

(attack damage carry)

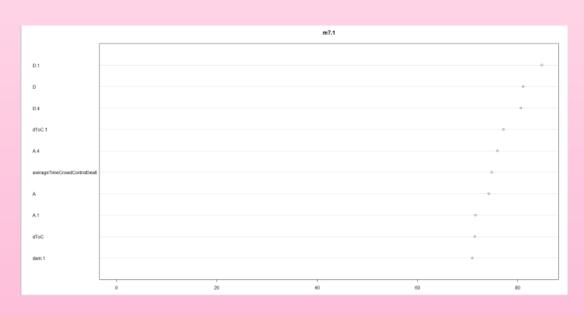
Support: Heal & Damage

Result: Rule for C5



- 1. Top Damage
- 2. Top Assist
- 3. Support Damage
- 4. ADC Damage

Result: Important variables for Randomforest



- 1. Support Damage
- 2. ADC Damage
- 3. Top Damage
- 4. Top Assist

Result: Important variables for XGBoost



- 1.Support DamageTakon
- 2. ADC HeroDamage
- 3. Middle HeroDamage
- 4. Jungle Assist

Conclusion

So far, we get some models with prediction rate nearly 60%. Some important variables are selected to help player have better understand for this game.

Variable selection and model parameter adjustment are our next concerns.

Thank You