

# Board games recommender

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### **Goal & Motivation**

- Develop a recommender system for board games
  - users who like to play board games and want to try new ones
- Possible monetization:
  - e-shop
  - referral links to e-shops

### Data

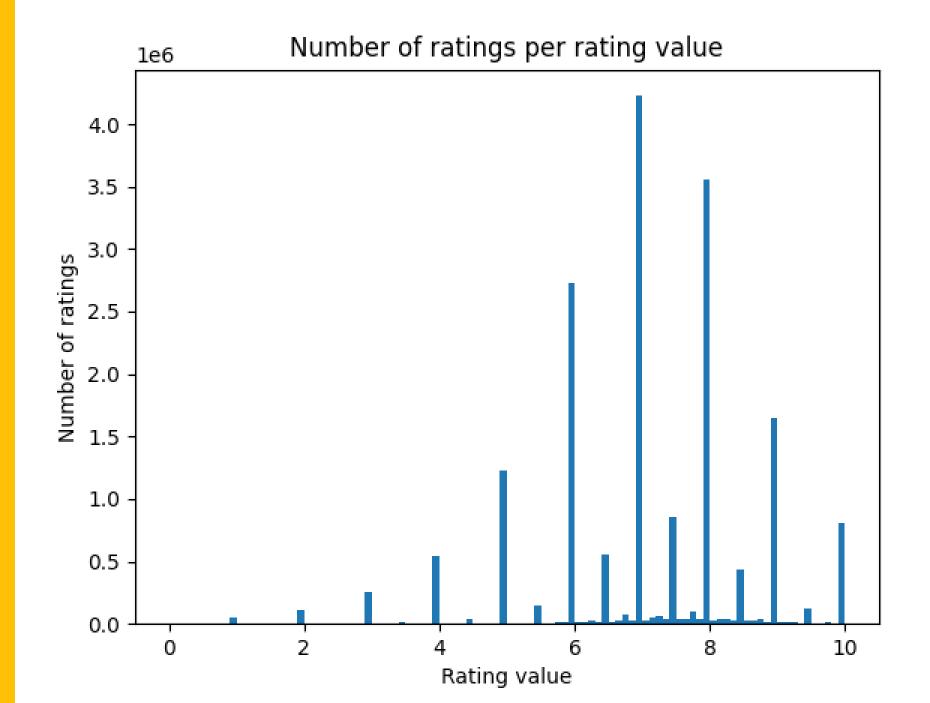
- from kaggle
- games, mechanics, subcategories, themes
- user\_ratings
  - BGGId BoardGameGeek game ID
  - Rating Raw rating given by user
  - Username User giving rating

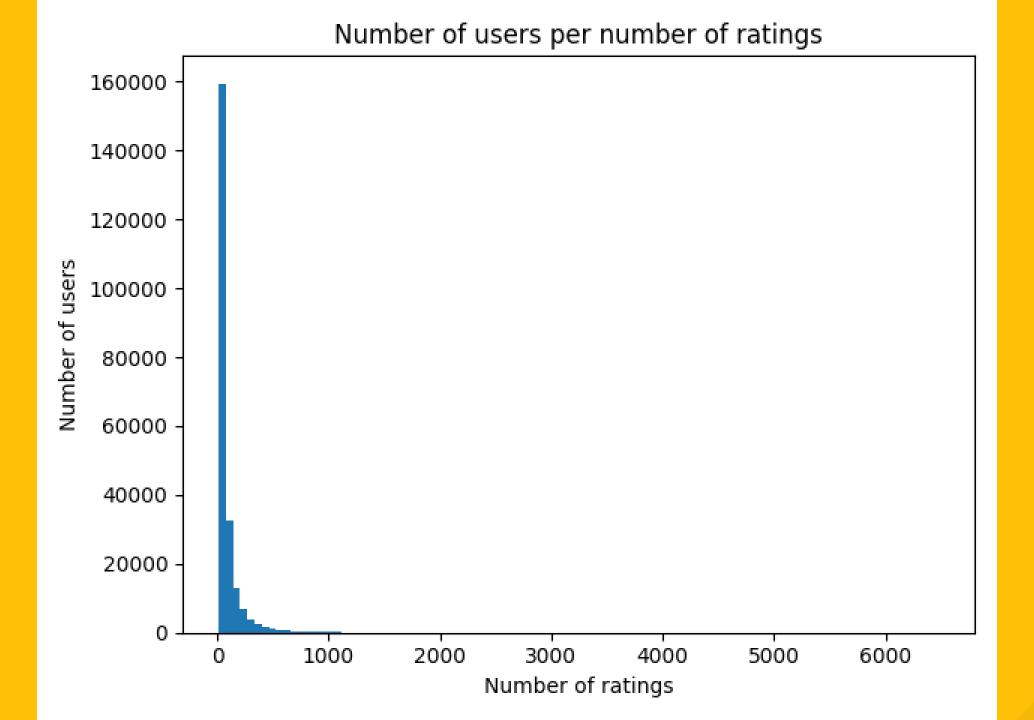
## Data preprocessing

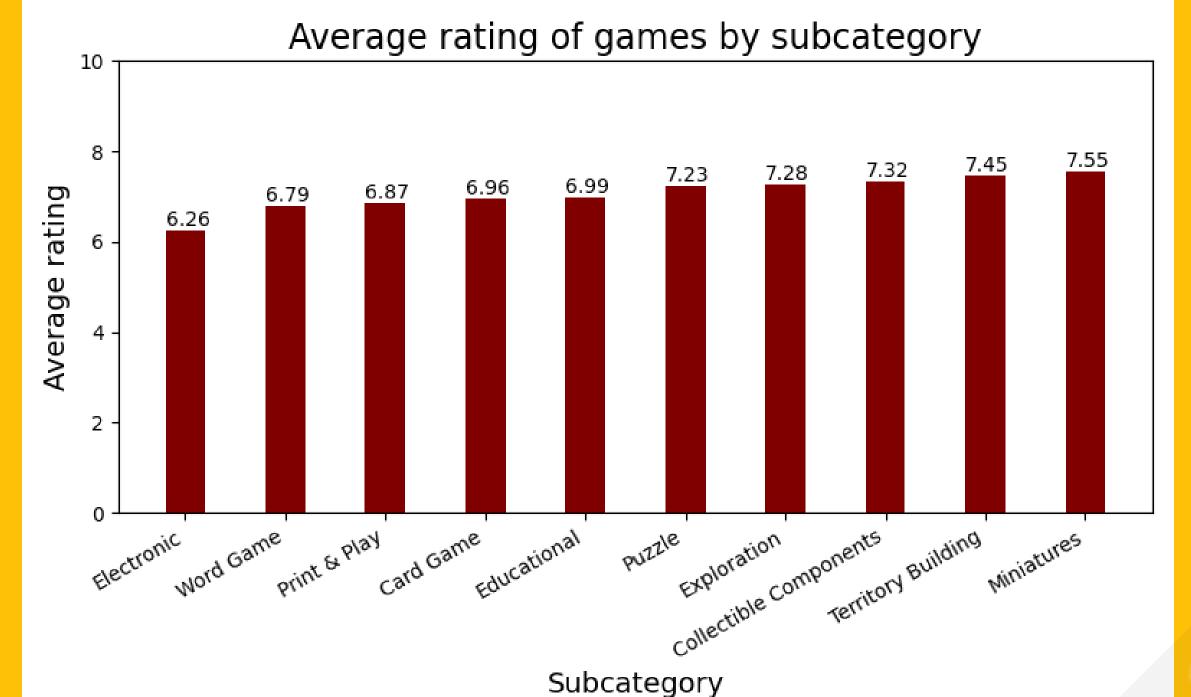
- removed duplicate ratings
- removed users with less than 10 ratings
- removed games with less than 10 ratings

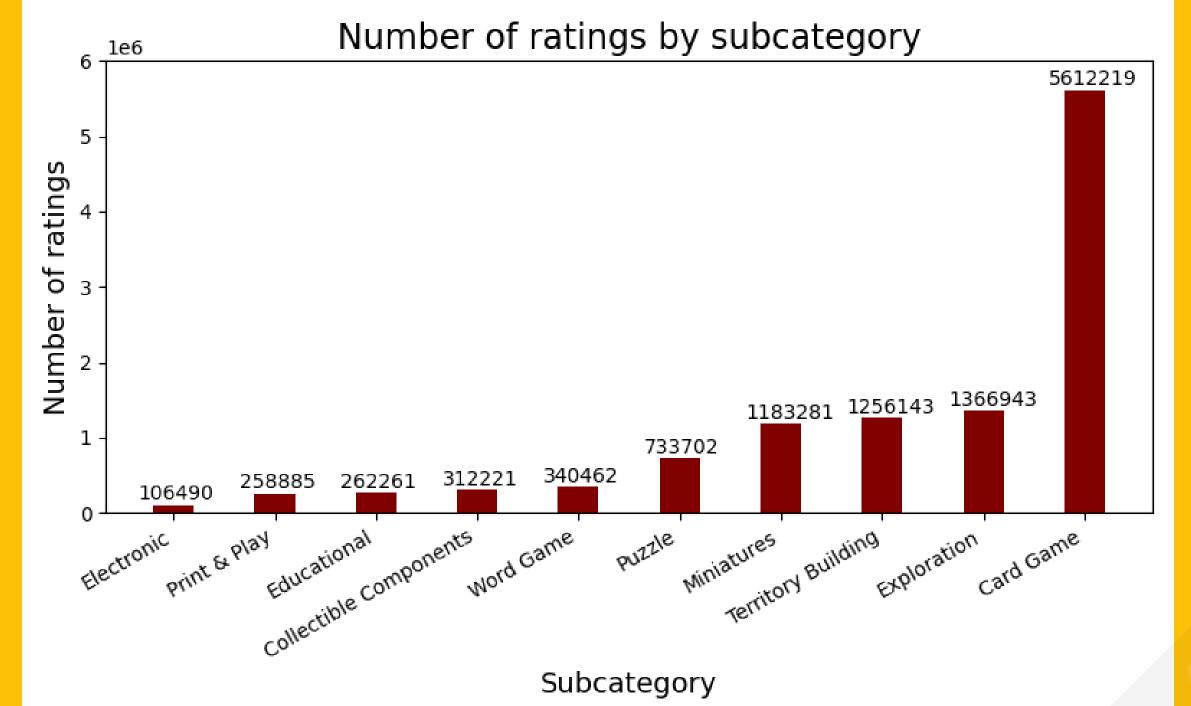
### Data analysis

- 18 340 284 user ratings
- 224 557 users, 21 919 games, 157 mechanics,
   217 themes, 10 subcategories
- density of user ratings matrix: 0.37%
- ratings per user:
  - average: 81.67
  - o median: 39
  - maximum: 6478
- average rating: **7.10** (range **[0, 10]**)









## **TF-IDF**

- most of relevant data are binary flags + some numerical values and Description
- pipeline:
  - convert binary flags and numerical values to text
  - concatenate with description
  - compute TF-IDF matrix
  - compute cosine similarities
  - choose rows with rated games
  - sort games based on similarity score

### **Problems**

- during development:
  - converting everything to text (so that library function could be used)
  - mapping index in matrix to index in database
- finished product:
  - reimplementations

### **Evaluation**

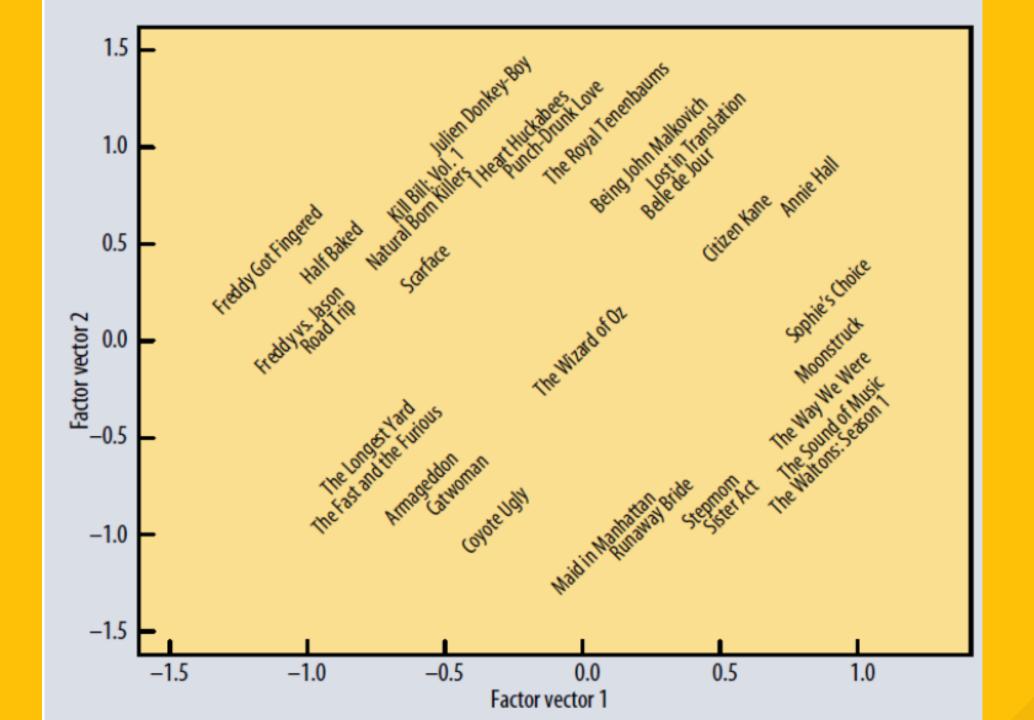
- no exact metric, just by feedback
- mostly positive feedback, but problems when game has too many reimplementations (those are very similar, so they get high score)
- examples of recommendations at the end

### Latent factors

- idea:
  - we are trying to model "taste" of users and "features" of items
  - matrices of user / item latent factors
- approach: minimize squared errors (+ regularization)
- method: stochastic / mini-batch gradient descent

$$\hat{r}_{ui} = q_i^T p_u$$

$$\min_{q,p} \sum_{(u,i)\in\mathcal{T}} (r_{ui} - q_i^T p_u)^2 + \lambda(||q_i||^2 + ||p_u||^2)$$



# Pipeline I

- split dataset into train, validation and test set
  - idea: all games should be present in all three datasets

# Pipeline II

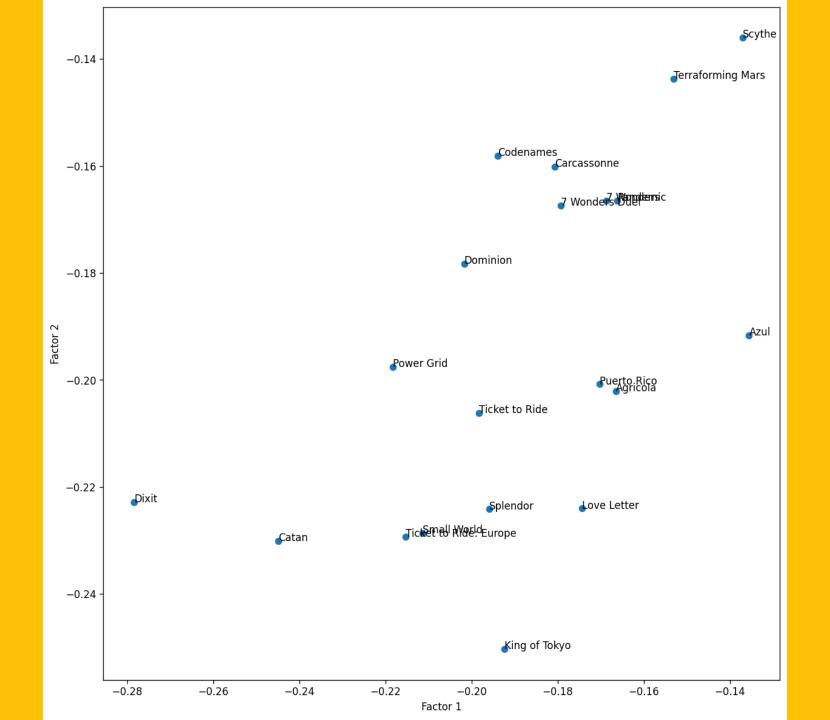
- use training set (user ratings) to train the model
  - adjust matrices of user / item factors using gradient descent
- use validation set to prevent overfitting
  - compute RMSE
  - apply early stopping if necessary
- use test set to evaluate the trained model (RMSE)

### Initial results I

- stochastic gradient descent too slow
  - necessary to reduce data
- adopt mini-batch gradient descent
  - allows to train on full data with more epochs / hyperparameter tuning

### **Initial results II**

- predictions not quite "reasonable" (no obvious pattern)
  - 2 latent factors → 2D plane (similar to PCA) → find features in similar / "opposite" clusters
  - compare RMSE of latent factors model with baselines



### RMSE comparison

- global mean: 1.530
- user mean: 1.376
- item mean: 1.316
- global mean + item/user bias: 1.230
- latent factors: 1.19
- latent factors with global effects: 1.19

### New user

- approximate user factors from ratings
- item factors matrix is constant
- least squares
- simple approach (systems of equations)
- examples of recommendations at the end

# **Experience report**

- necessary to implement gradient descent myself
- necessary to make mini-batch
- additional effort to confirm algorithms are implemented correctly
- computing recommendations for new user initially unclear

# Memory based CF

### • pipeline:

- ratings matrix (users × games) get users that rated the same games as me
- unrated game keep users who rated it
- find k most similar users to me
- get mean rating

### **Qualitative evaluation**

- only on local device (not integrated into the app)
- low amount of feedback testing in Postman
- not very intuitive, much novelty and unexpected recommendations
- stick to category RPGs recommend RPGs

```
http://localhost:8000/recommendations/item-based-cf
POST
                      Headers (9)
                                     Body Pre-request Script Tests Settings
Params
         Authorization
none form-data x-www-form-urlencoded raw binary GraphQL JSON v
        "offset": 0,
       "limit": 3,
       "ratings": [
       . . . . {
       ···· gameId": 823, // The Lord of the Rings
      "value": 10
       ....},
       · · · {
       .... gameId": 7467, // The Lord of the Rings: The Two Towers Board Game
     "value": 9
 11
 12
       ....},
 13
       · · · {
 14
       .... gameId": 77423, // The Lord of the Rings: The Card Game
       "value": 9
 15
       ....},
 17
       · · · · {
       ···· "gameId": 175755, ·//The Lord of the Rings: Journey to Mordor
 18
       ···· "value": 10
 19
 20
       • • • • }
 21
      | ]
 22
```

```
http://localhost:8000/recommendations/item-based-cf
POST
                                               Pre-request Script
Params
         Authorization
                      Headers (9)
                                      Body •
                                                                  Tests
                                                                          Settings
none form-data x-www-form-urlencoded raw binary GraphQL JSON v
        "offset": 0,
        "limit": 3,
        "ratings": [
         ...."gameId":.3955,.//.BANG
         --- "value": 10
        ···},
        ... "gameId": 463, // Magic: The Gathering
 11
        "value": 10
 12
        · · · },
 13
 14
         ... "gameId": 77423, .// The Lord of the Rings: The Card Game
         "value": 9
 15
        ··},
 17
 18
         ····gameId": 6366, ·// ·Dungeons ·& ·Dragons: ·The ·Fantasy ·Adventure ·Board ·Game
 19
         "value": 10
 21
 22
```

## **Encountered problems**

- performance not usable in real-time, not deployed
- implementation several new technologies (also programming on Windows)
- memory creation of ratings matrix on local device

# **User interface**

Rate games

Search all games by name

Q piranha



 $\times$ 



#### Piranha

From 2004

Rating: 5 / 10

Difficulty: 1 / 5





2 - 6 players



20 minutes

DETAILS

BUY

RATE



#### Piranha Panic

From 2005

Rating: 4 / 10

Difficulty: 1 / 5



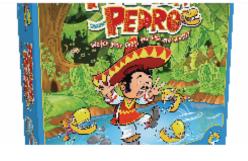


2 - 4 players



15 minutes

**DETAILS** BUY RATE



#### Piranha Pedro

From 2004

Rating: 6 / 10

Difficulty: 1 / 5





2 - 6 players



30 minutes

**DETAILS** 

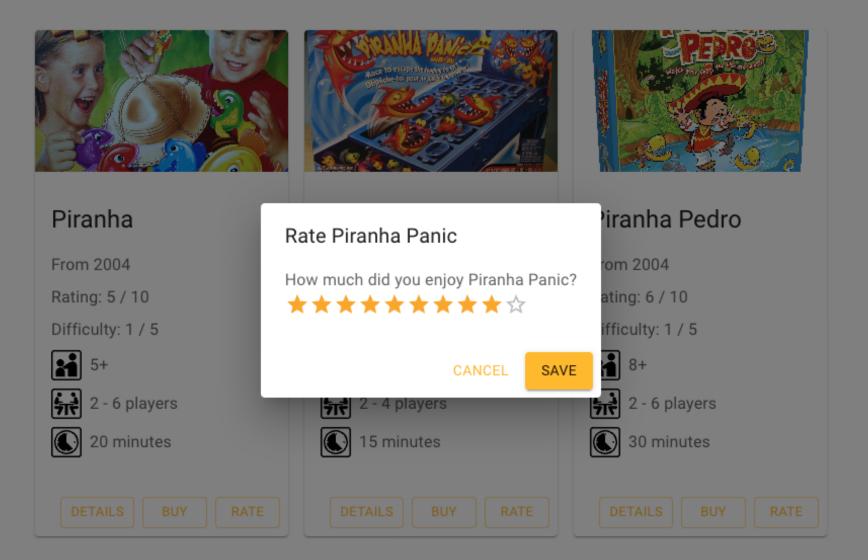
BUY

RATE

Rate games

Search all games by name

Q piranha



X

- Search my ratings by name





#### RATE GAMES

#### RESET RATINGS

Name	Image	Your rating	Rating given at
UNO			08/05/2023, 17:21:11
Catan	TRADE BUILD SETTLE.	******	08/05/2023, 17:20:33
Piranha Panic	Race to escape the fest yields to	*****	08/05/2023, 17:20:21
Rows per page: 3 ▼ 1–3 of 3 < >			

### TF-IDF



CATAN: 3D Edition

From 2021

Rating: 7 / 10

Difficulty: 2 / 5





3 - 4 players



120 minutes

DETAILS

BUY

RATE



Catan: 25th Anniversary Edition

From 2020

Rating: 7 / 10

Difficulty: 2 / 5





3 - 6 players



120 minutes

DETAILS

BUY

RATE



Catan: 25 Jahre Jubiläums-Edition

From 2020

Rating: 7 / 10

Difficulty: 0 / 5





3 - 4 players



120 minutes

DETAILS

BUY

RATE

### Latent factors





From 2017

Rating: 8 / 10

Difficulty: 1 / 5





**र्नारे** 1 - 4 players



90 minutes

DETAILS

BUY

RATE



#### Magi Kitchen

From 2014

Rating: 6 / 10

Difficulty: 1 / 5





4 - 8 players



15 minutes

DETAILS

RATE

BUY



#### Celestial Rainbows

From 2013

Rating: 5 / 10

Difficulty: 1 / 5





2 - 8 players



15 minutes

DETAILS

BUY

RATE

### Top-rated







#### Gloomhaven

From 2017

Rating: 8 / 10

Difficulty: 3 / 5





1 - 4 players



120 minutes

DETAILS

BUY

RATE

#### Pandemic Legacy: Season 1

From 2015

Rating: 8 / 10

Difficulty: 2 / 5





2 - 4 players



60 minutes

DETAILS

BUY

RATE

#### Brass: Birmingham

From 2018

Rating: 8 / 10

Difficulty: 3 / 5



14+



2 - 4 players



120 minutes

DETAILS

BUY

RATE

37

### Most rated







#### Pandemic

From 2008

Rating: 7 / 10

Difficulty: 2 / 5





2 - 4 players



45 minutes

DETAILS

BUY

RATE

#### Carcassonne

From 2000

Rating: 7 / 10

Difficulty: 1 / 5





2 - 5 players



45 minutes

DETAILS

BUY

RATE

#### Catan

From 1995

Rating: 7 / 10

Difficulty: 2 / 5





3 - 4 players



120 minutes

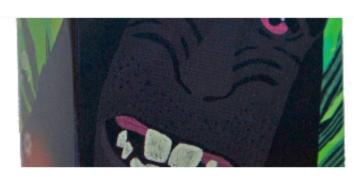
DETAILS

BUY

RATE

BUY RATE DETAILS RATE DETAILS BUY RATE

### Random picks



#### Monikers: Something Something

From 2016

Rating: 8 / 10

Difficulty: 1 / 5





4 - 20 players



60 minutes

**DETAILS** BUY RATE



#### Das Jagdspiel

From 1954

Rating: 5 / 10

Difficulty: 1 / 5





2 - 8 players



45 minutes

DETAILS BUY RATE



#### **Hot Spot**

From 1979

Rating: 5 / 10

Difficulty: 1 / 5





**र्वारे** 2 - 2 players



60 minutes

DETAILS

BUY

RATE

# Thank you



Try it out: https://pv254-board-gamesrecommender.vercel.app/

Code: https://github.com/petr7555/pv254-board-games-recommender