

# Crazy Arcade

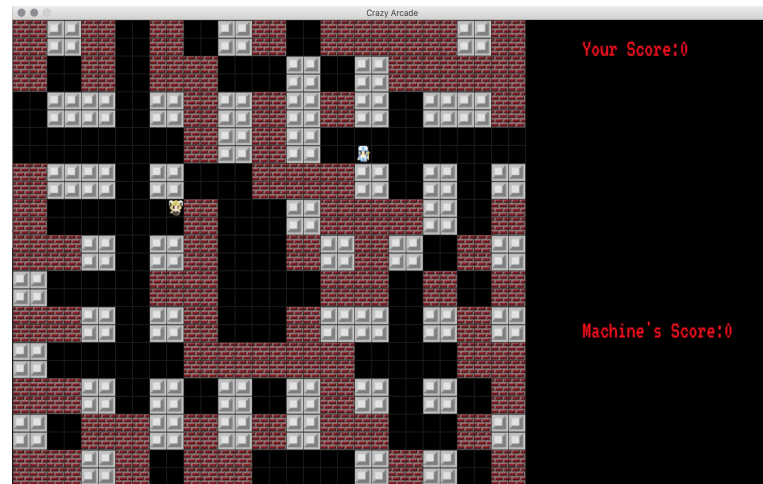
(Q版泡泡堂)

Lu Guo & Yuping Zang

# Agenda

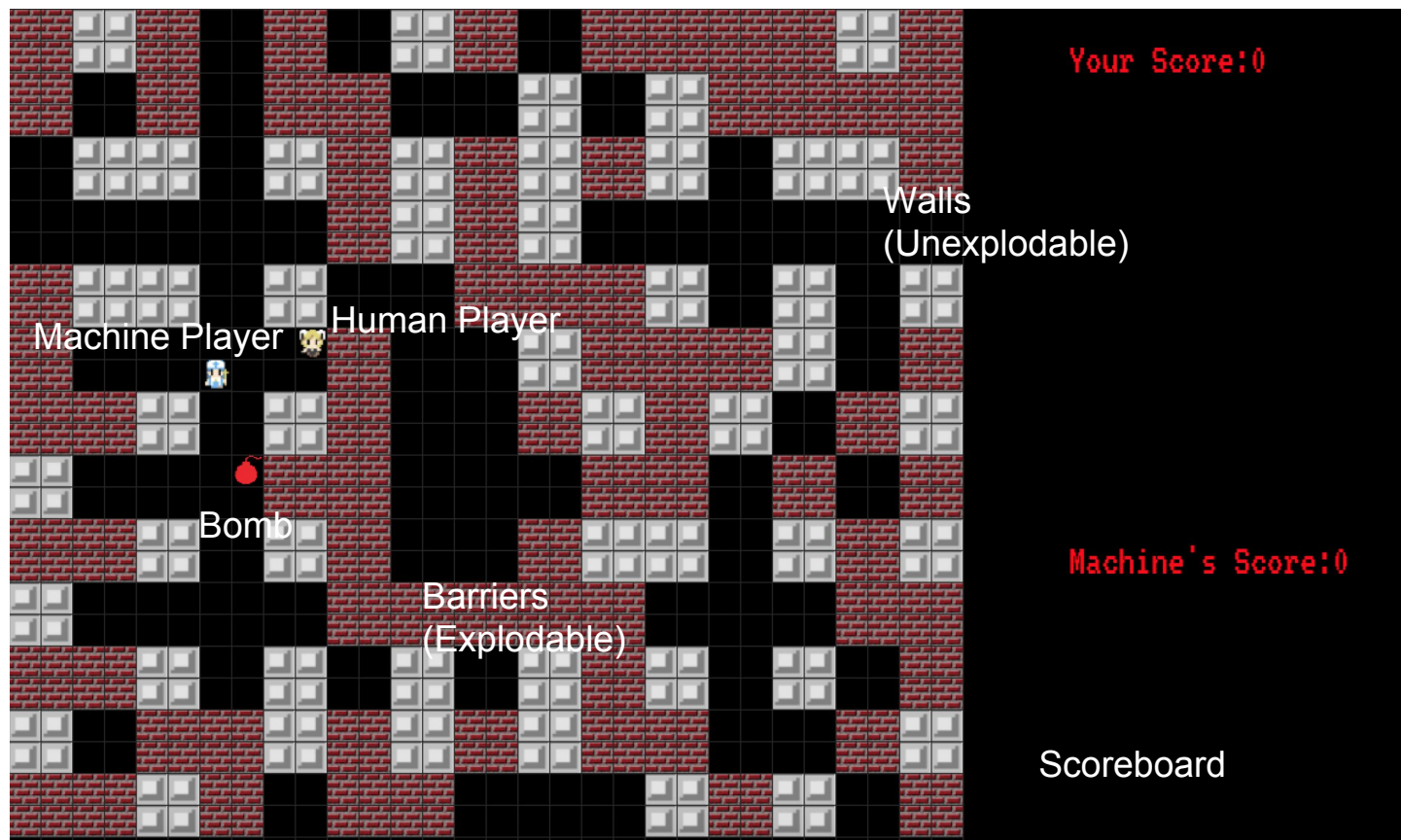
- Basic Game Rules
- Working Process Overview
- Several Highlights
- Reflection & Future Improvement

# Crazy Arcade: A Human V.S. Machine Game Based on the Chatting System



- Inspiration from Q版泡泡堂 on 4399.com
- Adapt it into a human-computer game with artificial intelligence

# Basic Game Rules



# Difficulty Levels: Different Parameters of Machine Simulation

-- how intelligent the machine is

## CRAZY ARCADE

Choose the Level

EASY

Maximum Simulation  
Trials = 100

NORMAL

Maximum Simulation  
Trials = 200

HARD

Maximum Simulation  
Trials = 300

# Mouse Control + Keyboard Control

- `pygame.mouse.get_pos()` : detect mouse position and button function to access different difficulty settings.
- Listen to keyboard inputs: control human player's movement event and bomb event.



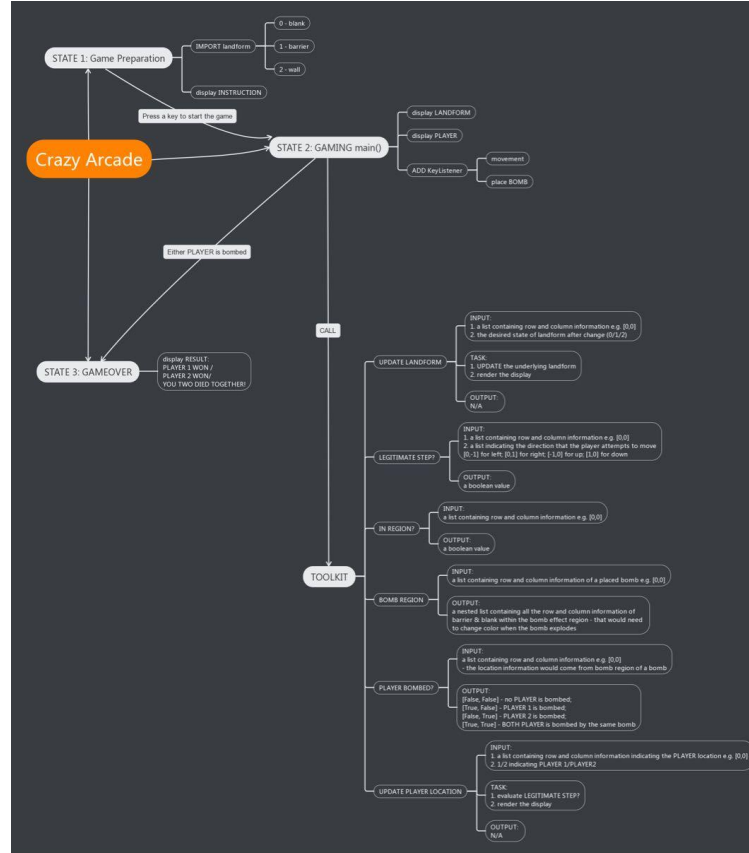
MOVEMENT

PLACE A BOMB

# Workflow Overview

1. Design state machine diagram & program flowchart
2. Implement game settings and human-controlled agent
3. Implement machine-generated agent with smart movement and bomb placement decisions based on random walk and simulation
4. Improve machine-generated agent's intelligence in reacting to dilemma of survival, attack from human agent, and urgent explosion...

# State Machine Diagram & Program Flowchart

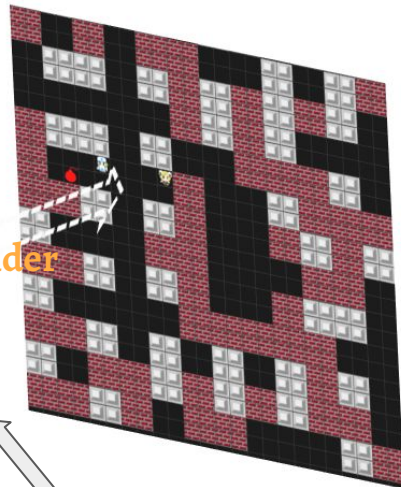
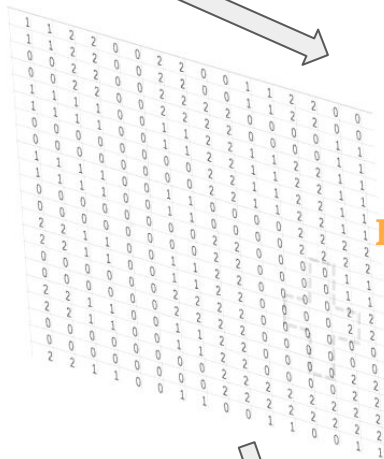




# Interaction between Underlying Landform Matrix & UI

**Human-Controlled Agent**

Keyboard input



Render

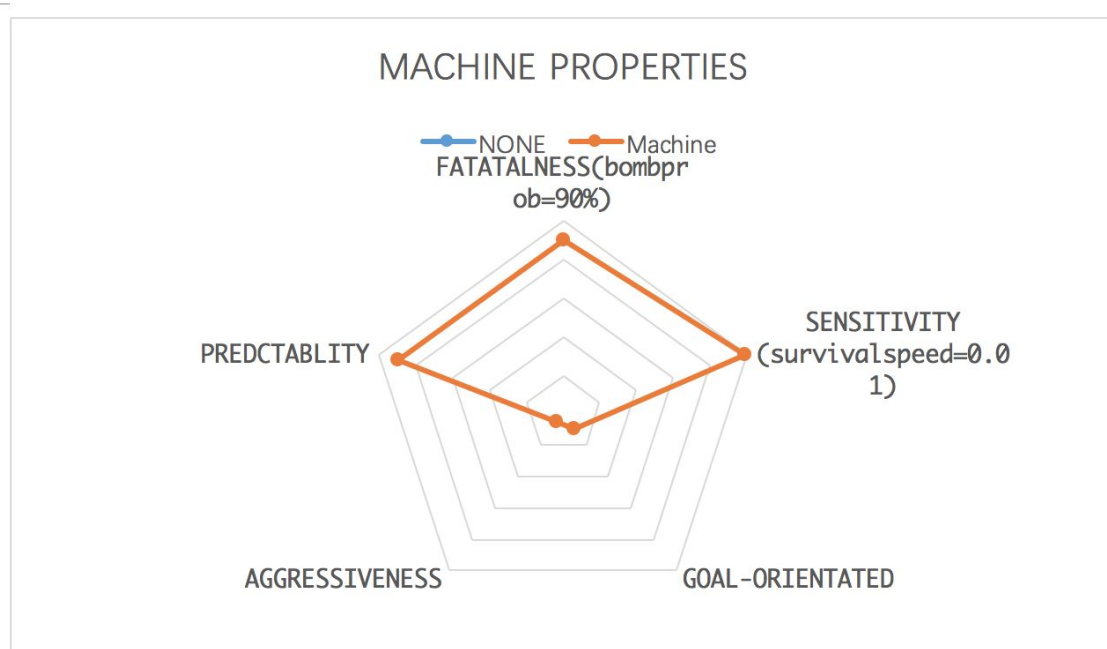
**Listen**

**Decision-Making**

**Machine-Generated Agent**

# Machine Property: Decision-Making Framework of Movement Event & Bomb Placement

- Movement Event: **Random Walk + Exit Danger Regions (Triggered by any activities that may lead to changes in landform)**
- Bomb Placement: **Certain Probability + Simulation Evaluation**

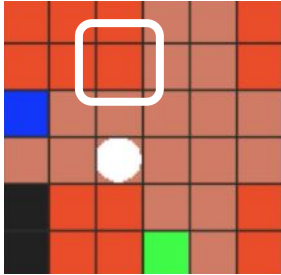


## Highlight 1: Machine's Urgent Response Mechanism

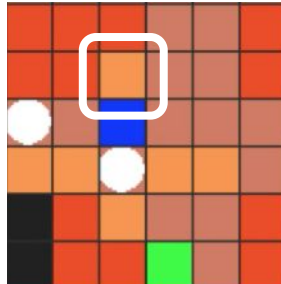
### KEY CHALLENGE:

Overlapped bombing events lead to chaotic dynamics in landform change.

1



2



### DEMONSTRATION:

Safe zone may transform into a potential danger zone after a bomb event.

## Highlight 1: Machine's Urgent Response Mechanism

- *InDilemma()*: Return a boolean value that indicates machine player's status if it is trapped in a danger zone with zero adjacent exit options among 4 directions.

-> Initialize a more aggressive and blind exploration in larger region to escape.

```
def InDilemma(index, MoreDangerZone = []):
    if InDangerZone(index) or (index in MoreDangerZone) or (index in BOMBSTACK):
        #ProbDirec = 0
        for i in DIRECTION:
            adjustedIndex = [index[0]+i[0], index[1]+i[1]]
            if legitMove(adjustedIndex) and not(InDangerZone(adjustedIndex)) and (adjustedIndex not in MoreDangerZone) and (adjustedIndex not in BOMBSTACK):
                #ProbDirec += 1
                return False
        #if ProbDirec == 0:
        return True
    return False
```

## Highlight 1: Machine's Urgent Response Mechanism

- *UrgentMove()*: Handle real-time reaction to possible attacks from human player.

-> If a bomb attack is detected, machine player would first follow a relatively safe trajectory sampled from 20 pre-set adjacent exit options, and then escape further if necessary after evaluation.

```
def UrgentMove(screen):
    global INURGENTMOVE
    global INSURVIVAL

    INURGENTMOVE = True
    # new
    while InDangerZone(PLAYERTWO) or (PLAYERTWO in BOMBSTACK):
        # new
        if InDilemma(PLAYERTWO):
            if not INSURVIVAL:
                Survive(screen)
                INURGENTMOVE = False
                return
            else:
                MachineMoveOneStep(screen)
                time.sleep(SURVIVALSPEED)
    INURGENTMOVE = False
```

## Highlight 1: Machine's Urgent Response Mechanism

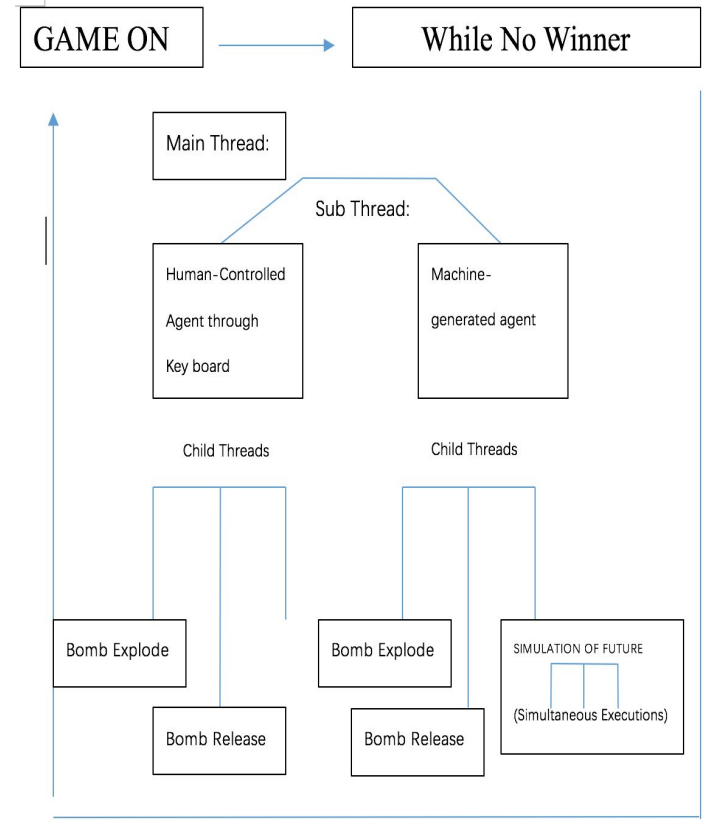
- *Survive()*: Perform a more aggressive (risk-neutral) and blind search in larger region in hope of wandering to a safe position.

-> Based on random walk. Limited by movement speed and maximum survival trials depend on difficulty levels.

```
def Survive(screen):  
    global INSURVIVAL  
  
    INSURVIVAL = True  
    for i in range(SURVIVALTRAIL):  
        MachineMoveOneStep(screen, False)  
        if not InDangerZone(PLAYERTWO):  
            INSURVIVAL = False  
            return  
        time.sleep(SURVIVALSPEED)  
    INSURVIVAL = False
```

## Highlight 2: Simultaneous Multi-Threading

- **Global Level: Human-controlled agent in main thread & machine-generated agent in sub thread**
- **Bomb Event: Bomb explode and bomb release as separate child threads (“sandbox”)**
- **Simulation Evaluation: Simultaneous execution in survival trials**
- **Integration into Chat System: As a subprocess.**



## Highlight 2: Simultaneous Multi-Threading

- **Global Level**

```
def running_game():  
    .....  
    _thread.start_new_thread(MachineMove,(screen,MACHINEMOVETIMELAG))
```

- **Simulation Evaluation**

```
def RunSimulation():  
    global SuicideTotal  
    from multiprocessing import Pool  
  
    THREADS = 10  
    msgs = [0.0 for x in range(SLTIME)]  
  
    with Pool(THREADS) as p:  
        results = p.map(task,msgs)  
  
    # Prob(Suicide) under random walk  
    return (results.count(1)/SLTIME) <= SUICIDETHRESHOLD
```

- **Bomb Event**

```
_thread.start_new_thread(machineexplode,(screen,bombIndex,3))  
_thread.start_new_thread(release,(screen,bombIndex,4))
```



# Highlight 3: Monte Carlo Simulation

## Each Iteration of Simulation

1. Create a virtual machine agent



2. Perform (*SLSTEP*)

3. Record if the virtual agent  
is trapped in dilemma through-  
out the simulation

\*

*SLTIME*  
times



Suicide Probability

= suicide accumulator / total simulation times



Compare with *Threshold*  
Make bomb placement decisions

## Highlight 3: Monte Carlo Simulation

```
def SimulateOneStep(initialLoc,VoidLoc,lastStep,ForbiddenZone=[]):
    Direc = DIRECTION[random.randint(0,3)]
    PotentialLoc = [VoidLoc[0]+Direc[0],VoidLoc[1]+Direc[1]]
    # evaluate 这个方向可否走, 并执行
    if lastStep:
        if legitMove(PotentialLoc) and not(InDangerZone(PotentialLoc)) and (PotentialLoc != initialLoc) and (PotentialLoc not in ForbiddenZone):
            VoidLoc[0] += Direc[0]
            VoidLoc[1] += Direc[1]
            return VoidLoc,False
        else:
            if InDilemma(VoidLoc,ForbiddenZone) or (VoidLoc == initialLoc):
                return VoidLoc, True
    else:
        if legitMove(PotentialLoc) and not(InDangerZone(PotentialLoc)) and (PotentialLoc != initialLoc):
            VoidLoc[0] += Direc[0]
            VoidLoc[1] += Direc[1]
            return VoidLoc,False
        else:
            # 在这一步random walk选择不动
            # 如果既动不了, 又处于dilemma, 就会死
            if InDilemma(VoidLoc,initialLoc):
                return VoidLoc, True
    return VoidLoc,False

# -----
def RunSimulation():
    global SuicideTotal
    from multiprocessing import Pool

    THREADS = 10
    msgs = [0.0 for x in range(SLTIME)]

    with Pool(THREADS) as p:
        results = p.map(task,msgs)

    # Prob(Suicide) under random walk
    return (results.count(1)/SLTIME) <= SUICIDETHRESHOLD

# -----
```

## Reflection & Future Improvement

- Defensive  $\longrightarrow$  Aggressive
  - through distance-computation, bomb placement strategies
- Random Walk  $\longrightarrow$  Reinforcement Learning
  - through Q-learning with  $\varepsilon$ -greedy exploration, more adaptive to given landform
- RNN
  - Realization: play with itself & learning from trials