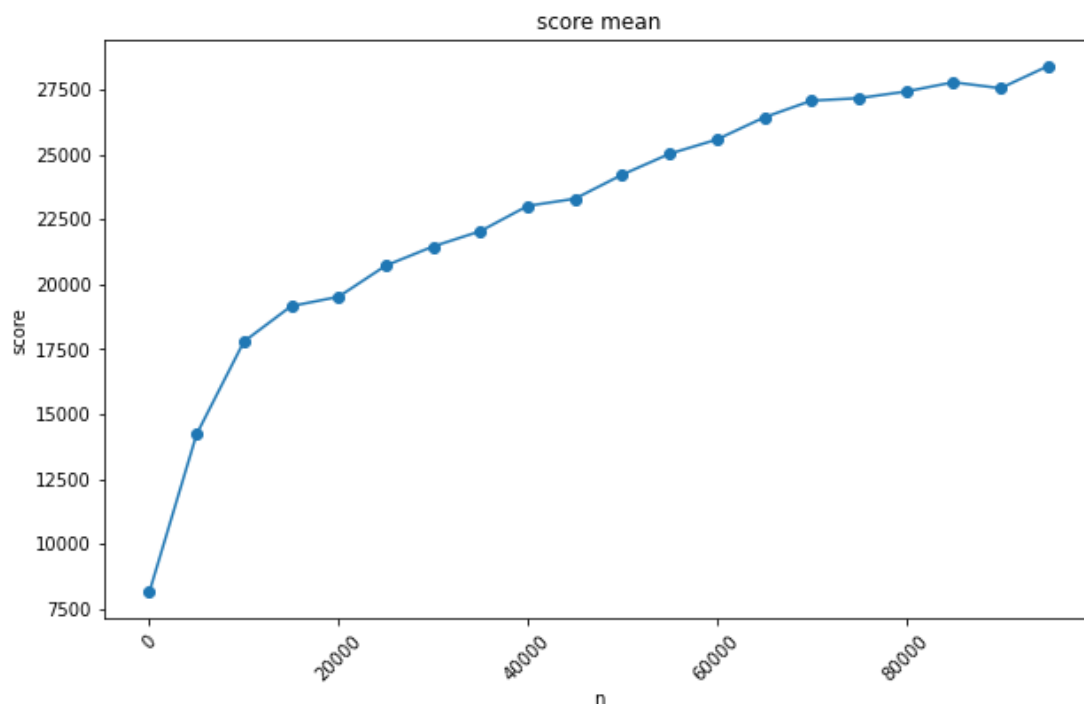
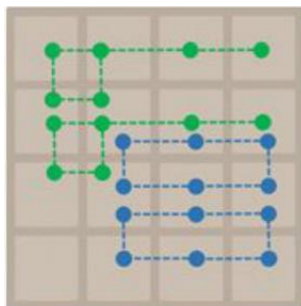


Performance:



Bonus:

- Describe the implementation and the usage of n -tuple network. (5%)
 n -tuple network 類似 cnn 取局部特徵，因為我們無法窮舉 16^{16} 所有可能性，所以改以 16^n 來代替，以 sample code 裡面就是取 6 位數 x 4 feature



盤面是可以 rotate 與 mirror 的，因此有 isomorphic(共構?)處理，一個 feature 會有 8 個相同性質，在計算 error 的時候也是 8 組 feature weight pair 做更新

- Explain the mechanism of TD(0). (5%)
 TD 跟 Q-learning 不同，value function 只針對 state 做評分

■ Update $V(\text{state})$, not $V(\text{after-state})$.

◆ That is, you need to use the information of $P(\text{popup tile 2}) = 0.9$ and $P(\text{popup tile 4}) = 0.1$ in your code.

```

function LEARN EVALUATION( $s, a, r, s', s''$ )
     $V(s) \leftarrow V(s) + \alpha(r + V(s'') - V(s))$ 

```

由這兩段我寫成如下：

```

void update_episode(std::vector<state>& path, float alpha = 0.1) const {
    // TODO
    float td_target = 0;
    // path是從頭到尾的操作sequence
    // pop_back從最後一個開始拿，會改變vector；back不會改變，純取不刪

    for (path.pop_back(); path.size(); path.pop_back()){
        state& move = path.back();
        // 不是很理解這個requirement: Update V(state), not V(after-state)
        // t = t 的 after state都還沒有考慮進popup，所以要使用 t+1 的before_state

        /**
         * s = before state, s' = after state, s'' = t+1 's before state
         * V(s) = V(s) + alpha (reward + V(s'') - V(s))
         */
        // td error = r_{t+1} + gamma * V(s_{t+1}) - V(s_t)
        // td error = td target - V(s_t)
        float td_error = td_target - estimate(move.before_state());

        // update return回來的已經是value function調整過後的
        // td target = r_{t+1} + gamma * V(s_{t+1})
        td_target = move.reward() + update(move.before_state(), alpha * td_error);
    }
}

```

我想說如果是考慮完 popup 後的盤面，應該都要取每個 state 的 before board 來計算($s_{t+1} - s_t$)，如果是 Q-learning 看 action 則用 after($s'_{t+1} - s'_t$)，但後者的表現似乎比前者好...可能是哪裡寫錯，不然 performance 有相當差距。

- Describe your implementation in detail including action selection and TD-backup diagram. (10%)

一個 feature 的 weight 在算時要考量共構的

```

virtual float estimate(const board& b) const {
    // TODO
    float value = 0;
    for (int i = 0; i < iso_last; i++)
    {
        size_t table_key = indexof(isomorphic[i], b);

        // operator[]已經被override，可以輸入table key而找到對應的weight

        value += operator[](table_key);
    }
    return value;
}

```

```

/**
 * update the value of a given board, and return its updated value
 * u代表 alpha*error，算出來的error要除8，表示feature與他的共構小夥伴共8個
 */
virtual float update(const board& b, float u) {
    // TODO
    float adjust = u / iso_last;
    float value = 0;
    for (int i = 0; i < iso_last; i++)
    {
        // size_t是unsigned int
        size_t table_key = indexof(isomorphic[i], b);
        // update weight
        operator[](table_key) += adjust;

        //return esti value that the weight is changed
        value += operator[](table_key);
    }
    return value;
}

```

```

size_t indexof(const std::vector<int>& patt, const board& b) const {
    // TODO
    // 從000000~ffffff都有對應一組weight，所以會有key value pair(table for lookup)來記錄，這裡就是input為board與board idx 而可以output對應的key
    size_t table_key = 0;
    /**
     * input
     * b = 0x 4312 7521 8653 2731
     * patt = {0, 1, 2, 3, 4, 5}
     */
    for (size_t i = 0; i < patt.size(); i++)
        table_key |= b.at(patt[i]) << (4 * i);
    return table_key;
    // 0x000000
    // 0x000001
    // 0x000031
    // 0x000731
    // 0x002731
    // 0x032731
    // 0x532731
}

```

這裡其實就是 `argmax_a`，用 `assign` 來檢查動作是否 legal

```

state select_best_move(const board& b) const {
    state after[4] = { 0, 1, 2, 3 }; // up, right, down, left
    // initialize 4 state obj for constructor input(opcode = 0,1,2,3)代表上下左右四個動作的state
    //每個state會有 board: before_state, after_state，before就是當前board，after存做完動作後的board
    state* best = after;
    for (state* move = after; move != after + 4; move++) {
        //檢查上下左右的move是否為legal，亦即score是否不為-1，
        //assing就可以set before state跟after state兩個board 與 score
        if (move->assign(b)) {
            // TODO

            // score跟esti分別代表遊戲給的reward R_t與模型如何評估盤面好壞的V(S_t)
            // move->set_value(move->reward() + estimate(move->after_state()));

            //TODO 考慮popup的element 2(0.9) 4(0.1)
            std::vector<board> all_possible_afterstate;
            std::vector<float> all_possible_afterstate_prob;
            std::vector<int> empty_idx;

```

`assign` 完之後的 `afterstate` 還未考量到 `popup` 後的所有可能，需要窮舉所有可能會長 2、4 的位置，以 `empty_idx` 紀錄

```

for(int i = 0; i < 16 ; i++){
    if (move->after_state().at(i) == 0)
    {
        empty_idx.push_back(i);
    }
}

for(int idx: empty_idx){
    board tmp = move->after_state();
    tmp.set(idx, 2);
    all_possible_afterstate.push_back(tmp);
    all_possible_afterstate_prob.push_back(0.9);
}
for (int idx : empty_idx)
{
    board tmp = move->after_state();
    tmp.set(idx, 4);
    all_possible_afterstate.push_back(tmp);
    all_possible_afterstate_prob.push_back(0.1);
}

```

嘗試做這段 pseudo code

```

function EVALUATE( $s, a$ )
     $s', r \leftarrow \text{COMPUTE AFTERSTATE}(s, a)$ 
     $S'' \leftarrow \text{ALL POSSIBLE NEXT STATES}(s')$ 
    return  $r + \sum_{s'' \in S''} P(s, a, s'') V(s'')$ 

```

```

float sum_value = 0;
for (int i = 0; i < all_possible_afterstate.size(); i++)
{
    // 2's weight: 0.9/13 ; 4's weight:0.1/13
    //std::cout << "board size:" << all_possible_afterstate.size() << std::endl;
    sum_value += estimate(all_possible_afterstate[i]) * (all_possible_afterstate_prob[i] / (all_possible_afterstate.size()/2));
}
//std::cout << "sum value:" << sum_value << std::endl;
// r + sum_s'' in S'' P(s,a,s'')V(s'')
move->set_value(move->reward() + sum_value);

if (move->value() > best->value())
    best = move;
} else {
    move->set_value(-std::numeric_limits<float>::max());
}
}
debug << "test " << *move;

```

筆記

int64 代表一個 board 的盤面

```
* 64-bit bitboard implementation for 2048
*
* index:
*  0  1  2  3
*  4  5  6  7
*  8  9 10 11
* 12 13 14 15
* note that the 64-bit value is little endian
* therefore a board with raw value 0x 4312 7521 8653 2731 ull would be
* +-----+
* |      2      8   128    4|
* |      8     32    64   256|
* |      2      4    32   128|
* |      4      2     8    16|
* +-----+
```

ull 是 unsigned long long

4312752186532731 的解讀方式是由左往右，所以 index 其實是 15, 14, 13, ..., 1, 0

```
/**
 * get a 16-bit row
 */
int fetch(int i) const { return ((raw >> (i << 4)) & 0xffff); }
```

$0 \leq i \leq 3$

先往左 $i*4$ 格之後，再取最後四位

4312 7521 8653 2731 如果我要取 $i = 0$

$\text{raw} \gg (0 \ll 4)$ ， $0 \ll 4$ 就還是 0，即 $\text{raw} \gg 0$ 還是 raw，然後和 0xffff 做 and

4312 7521 8653 2731

0000 0000 0000 ffff

0000 0000 0000 2731

```
/**
 * set a 16-bit row
 */
void place(int i, int r) { raw = (raw & ~(0xffffULL << (i << 4))) |
(uint64_t(r & 0xffff) << (i << 4)); }
```

除了要 set 的那行以外其他不變 所以 取 `or row = row|new`

```
/**
 * get a 4-bit tile
 */
int at(int i) const { return (row >> (i << 2)) & 0x0f; }
```

因為每個 tile 是 4 個 bit 所以是 `<< 2` 之後取 `0x0f(0x0000 0000 0000 1111)`

`0 << 2 = 0`

`1 << 2 = 4`

`2 << 2 = 8`

`3 << 2 = 12`

Isomorphic 共構，為了處理 rotate 與 mirror

```
/**
 * isomorphic patterns can be calculated by board
 *
 * take pattern { 0, 1, 2, 3 } as an example
 * apply the pattern to the original board (left), we will get
0x1372
 * if we apply the pattern to the clockwise rotated board
(right), we will get 0x2131,
 * which is the same as applying pattern { 12, 8, 4, 0 } to the
original board
 * { 0, 1, 2, 3 } and { 12, 8, 4, 0 } are isomorphic patterns
 * +-----+ +-----+
 * | 2 8 128 4| | 4 2 8 2|
 * | 8 32 64 256| | 2 4 32 8|
 * | 2 4 32 128| ----> | 8 32 64 128|
 * | 4 2 8 16| | 16 128 256 4|
 * +-----+ +-----+
 *
 * therefore if we make a board whose value is
0xfedcba9876543210ull (the same as index)
 * we would be able to use the above method to calculate its 8
isomorphisms
 *
 * 簡言之就是一個 feature 會有 8 個同性質的，這邊就一併處理
 * 如果我想抓 pattern { 0, 1, 2, 3 }，但實際上{ 12, 8, 4, 0 }就是旋
轉之後的{ 0, 1, 2, 3 }
```

```
* 旋轉 4 個 mirror 四個 一共 8 個
*/
for (int i = 0; i < 8; i++) {
    board idx = 0xfedcba9876543210ull;
    //處理 mirror
    if (i >= 4) idx.mirror();
    //處理 rotate
    idx.rotate(i);
    // p = pattern({ 0, 1, 2, 3, 4, 5 })
    for (int t : p) {
        isomorphic[i].push_back(idx.at(t));
        // { 0, 1, 2, 3, 4, 5 }
        // { 3, 7, 11, 15, 2, 6 }
        // { 15, 14, 13, 12, 10, 11 }
        // { 0, 4, 8, 12, 9, 12 }
        // 懶得列 mirror 了
    }
}
```

indexof

找到 feature 對應的 index，也就是這件事

64	● ⁰	8	4
128	2● ¹		2
2	8● ²		2
128	● ³		

0123	weight
0000	3.04
0001	-3.90
0002	-2.14
⋮	⋮
0010	5.89
⋮	⋮
0130	-2.01
⋮	⋮

```
size_t indexof(const std::vector<int>& patt, const board& b) const {
    // TODO
    // 從 000000~ffffff 都有對應一組 weight，所以會有 key value
    pair(table for lookup)來記錄，這裡就是 input 為 board 與 board idx 而可以
    output 對應的 key

    size_t table_key = 0;
    /**
     * input
     * b = 0x 4312 7521 8653 2731
     * patt = {0, 1, 2, 3, 4, 5}
     */
    for (size_t i = 0; i < patt.size(); i++)
        table_key |= b.at(patt[i]) << (4 * i);
    return table_key;

    // 0x000000
    // 0x000001
    // 0x000031
    // 0x000731
    // 0x002731
    // 0x032731
    // 0x532731

}
```


estimate

```
/**
 * estimate the value of a given board
 */
virtual float estimate(const board& b) const {
    // TODO
    float value = 0;
    for (int i = 0; i < iso_last; i++)
    {
        size_t index = indexof(isomorphic[i], b);

        // operator[]已經被 override，可以輸入 table key 而找到對應的
weight
        // 把 board 上有出現的 feature 對應的 weight sum 起來
        value += operator[](index);
    }
    return value;
}
```