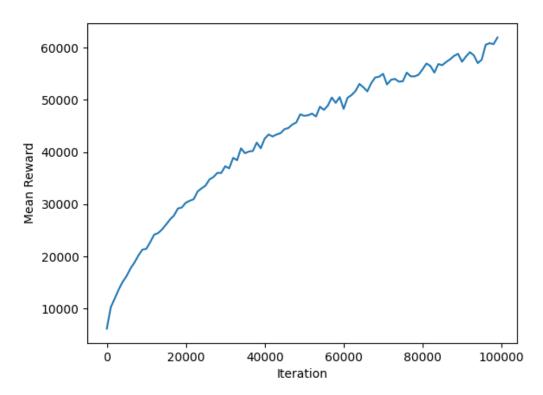
## **RL Topic HW1**

## 1. Training Curve



- 2. Implementation and usage of n-tuple network
  - a. Usage: n-tuple network would add all isomorphic pattern of each n-tuple of the current board to estimate the value.
  - b. Implementation: First, add all value of the n-tuples, a 'feat' is a n-tuple pattern.

and then query the summation of all 8 isomorphic of current n-tuple pattern.

the 'index\_of' would encode the numbers on board according to an isomorphic into a (4 \* n) bits integer.

- 3. Mechanism of TD(0)
  - a. It would update the current value by an single step reward and next estimation, without the need of calculated the discounted total return.
- 4. Implementation detail
  - a. Action selection:

For each valid action (valid means it would actually move the numbers), it will firstly find the empty spaces, then calculate the expectation of the board from generating a number (90% 2, 10% 4) on this empty space.

After summing up all the expected values, take average of all the values, since all space would have a same probability to be selected to generate a number. Then we add the average with the action reward, then it would be the value regarding the action. Afterall, we select the action with the largest action value.

```
      718
      total_value /= (float)total_space;

      719
      total_value += (float)move->reward();

      720
      move->set_value( v: total_value);

      721
      move->set_value() > best->value())

      722
      if (move->value() > best->value())

      724
      best = move;
```

b. TD-backup

We can get the TD-update by the formulation:

$$V(s) \leftarrow V(s) + \alpha (r + V(s'') - V(s))$$

And the value of r + V(s'') - V(s) can be easily calculated in a reversed way.

Since the update function would return the updated value, so when we have a path of  $s_0 \to s_1 \to s_2 \to s_x$ , then we can update  $s_2$  by  $\alpha(r_2 + V(s_x)[= 0] - V(s_2))$ , then update  $s_1$  by  $\alpha(r_1 + V(s_2)[= update(s_2)] - V(s_1))$ .

```
void update_episode(std::vector<state>& path, float alpha = 0.1) const {

// rood
float next_est = 0.0;
for(auto it1:reverse_iterator<...> = ++path.rbegin(); it1 != path.rend(); ++it1) {

float err = (float)it1->reward() + next_est - estimate( b: it1->before_state());

next_est = update( b: it1->before_state(), u: alpha * err);
}

754
}
```