Qishi Quiz 4 Solution

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1 Math

Question 1

The number 1, n-1, n has equal probability to be taken first. The last 2 persons get correct seat if and only if the seat 1 is taken before n-1 and n. So the probability is 1/3 when $n \ge 3$.

Question 2

There are 5 people in total, and we are going to calculate the total number of different arrangements in 5 days. Assume we have already find 5 pairs , which satisfies the requirement. Then any permutation of these 5 pairs gives one valid arrangement. The number of all possible permutation for requirement satisfying pairs is 5!.

So we only need to calculate the number of requirement satisfying pairs. We treat girl as node, and if two girl guard the store at the same day, we draw one edge between these two nodes. Then finding the number of requirement satisfying pairs is equivalent to find the number of all possible circle connecting these 5 nodes. This guarantees that each girl works twice a week.

WLOG, start from 1, so 4 choices for the next, then 3 choices, ... Each path is the same in terms of the cycle with the reversed path. So the number of all possible circle is $\frac{1}{2}4!$.

So the number of total arrangement is

$$\frac{1}{2}4! \cdot 5!$$

Question 3

Similar to the previous question, the only difference is that now we can have two cycles, with 3 nodes in one cycle and 4 nodes in the other. This case does not happen in the previous question because there were 5 nodes and no repeated pairs allowed.

To connect all nodes by circle, there are two possibility. One is a circle containing all 7 nodes, one is two circle containing 3 nodes and 4 nodes separately. So in total, the number of all possible circle connecting these 7 nodes is

$$\frac{1}{2}6! + \binom{7}{3} \cdot \frac{1}{2}2! \cdot \frac{1}{2}3!$$

So in total, the number of all possible arrangement is

$$7!(\frac{1}{2}6! + \binom{7}{3} \cdot \frac{1}{2}2! \cdot \frac{1}{2}3!)$$

Assume the stock is driven by a Geometric Brownian Motion,

$$S_{t_1} = S_0 \exp((\mu - \frac{\sigma_1^2}{2})t_1 + \sigma_1 W_{t_1})$$

$$S_{t_1} = S_0 \exp((\mu - \frac{\sigma_2^2}{2})t_2 + \sigma_1 W_{t_2})$$

So we have

$$corr(S_{t_1}, S_{t_2}) = corr(\exp(\sigma_1 W_{t_1}), \exp(\sigma_2 W_{t_2}))$$

$$= \frac{cov(\exp(\sigma_1 W_{t_1}), \exp(\sigma_2 W_{t_2}))}{\sqrt{var(\exp(\sigma_1 W_{t_1}))}\sqrt{var(\exp(\sigma_2 W_{t_2}))}}$$

Notice

$$var(\exp(\sigma W_t)) = \mathbb{E}[\exp(2\sigma W_t)] - \mathbb{E}[\exp(\sigma W_t)]^2$$
$$= \exp(2\sigma^2 t) - \exp(\sigma^2 t)$$

and

$$\begin{aligned} cov(\exp(\sigma_{1}W_{t_{1}}), \exp(\sigma_{2}W_{t_{2}})) &= \mathbb{E}[\exp(\sigma_{1}W_{t_{1}})\exp(\sigma_{2}W_{t_{2}})] - \mathbb{E}[\exp(\sigma_{1}W_{t_{1}})]\mathbb{E}[\exp(\sigma_{2}W_{t_{2}})] \\ &= \mathbb{E}[\exp((\sigma_{1} + \sigma_{2})W_{t_{1}})\exp(\sigma_{2}(W_{t_{2}} - W_{t_{1}})] - \mathbb{E}[\exp(\sigma_{1}W_{t_{1}})]\mathbb{E}[\exp(\sigma_{2}W_{t_{2}})] \\ &= \mathbb{E}[\exp((\sigma_{1} + \sigma_{2})W_{t_{1}})]\mathbb{E}[\exp(\sigma_{2}(W_{t_{2}} - W_{t_{1}})] - \mathbb{E}[\exp(\sigma_{1}W_{t_{1}})]\mathbb{E}[\exp(\sigma_{2}W_{t_{2}})] \\ &= \exp(\frac{(\sigma_{1} + \sigma_{2})^{2}}{2}t_{1})\exp(\frac{\sigma_{2}^{2}}{2}(t_{2} - t_{1})) - \exp(\frac{\sigma_{1}^{2}}{2}t_{1} + \frac{\sigma_{2}^{2}}{2}t_{2}) \end{aligned}$$

To conclude,

$$corr(S_{t_1}, S_{t_2}) = \frac{\exp(\sigma_1 \sigma_2 t_1) - 1}{\sqrt{\exp(\sigma_1^2 t_1) - 1} \sqrt{\exp(\sigma_2^2 t_2) - 1}}$$

Question 5

Only when N=0 and $N=1,\,W_t^N$ is a martingale. When N=3, we have

$$\mathbb{E}[W_t^3 | W_s^3] = \mathbb{E}[(W_s + W_t - W_s)^2 | W_s^3]$$

= $W_s^3 + 3W_s(t - s)$

So $\{W_t^3\}_t$ is a not a martingale.

Under risk neutral measure Q, we have

$$\frac{dS_t}{S_t} = rdt + \sigma dW_t$$

So we have

$$S_t = S_0 \exp((r - \frac{\sigma^2}{2})t + \sigma W_t)$$

For the option price,

$$\mathbb{E}\left[\frac{1}{S_T}\right] = \mathbb{E}\left[\frac{1}{S_0} \exp(-(r - \frac{\sigma^2}{2})t - \sigma W_t)\right]$$
$$= \frac{1}{S_0} \exp(\sigma^2 T - rT)$$

Question 7

Consider the problem to be solved by the numerical algorithm as a function f mapping the data x to the solution y. The result of the algorithm, say y^* , will usually deviate from the "true" solution y. The main causes of error are round-off error and truncation error. The forward error of the algorithm is the difference between the result and the solution; in this case, $\Delta y = y^* - y$. The backward error is the smallest Δx such that $f(x + \Delta x) = y^*$; in other words, the backward error tells us what problem the algorithm actually solved. The forward and backward error are related by the condition number: the forward error is at most as big in magnitude as the condition number multiplied by the magnitude of the backward error.

Explicit methods calculate the state of a system at a later time from the state of the system at the current time, while implicit methods find a solution by solving an equation involving both the current state of the system and the later one. Mathematically, if Y(t) is the current system state and $Y(t + \Delta t)$ is the state at the later time Δt is a small time step, then for an explicit method

$$Y(t + \Delta t) = F(Y(t))$$

while for an implicit method one solves an equation

$$G(Y(t), Y(t + \Delta t)) = 0$$

Question 8

Yes, there exits a moment the success rate is exactly 0.5. Consider two counts: the count of success X and the count of failure Y. Each time we can only increase one of them by 1. From X < Y to X > Y there must be a moment where X = Y. Since both X, Y are integers and we can add 1 each time.

We should select the strategy with higher P&L.

2 Programming

return False

return True

Question 10

```
// line function pass point \boldsymbol{x} and \boldsymbol{y}
def line(x, y):
  return lambda z: (x[0] - y[0]) * (z[1] - y[1]) - (x[1] - y[1]) * (z[0] - y[0])
def inside_triangle(x, y, z, w):
  f = line(x, y)
  t = f(z) * f(w)
  if t <= 0:
    return False
  f = line(x, z)
  t = f(y) * f(w)
  if t <= 0:
    return False
  f = line(y, z)
  t = f(x) * f(w)
  if t <= 0:
    return False
  return True
Question 11
def valid_palindrome(s):
  if not s:
    return True
  i, j = 0, len(s) - 1
  while i < j:
    if s[i] == s[j]:
      i += 1
      j -= 1
    else:
```

```
def longest_decrease_subarray(1):
    if not 1:
        return 0

length = []
for ind, val in enumerate(1):
    if ind == 0:
        length.append(1)
        continue

if val <= l[ind - 1]:
        length.append(length[-1] + 1)
    else:
        length.append(0)

return max(length)</pre>
```

Question 13

```
Python code:
```

```
from collections import defaultdict
class Solution:
    Oparam board: the sudoku puzzle
    Oreturn: nothing
    def __init__(self, n=9):
        self.n = n
        self.n_sqrt = int(n ** 0.5)
        self.rows = [defaultdict(int) for _ in range(self.n)]
        self.cols = [defaultdict(int) for _ in range(self.n)]
        self.boxes = [defaultdict(int) for _ in range(self.n)]
        self.success= False
   def _get_box_id(self, row, col):
        return int((row // self.n_sqrt) * self.n_sqrt + col // self.n_sqrt)
   def _fill(self, board, row, col, val):
        if val != 0:
            self.rows[row][val] = 1
            self.cols[col][val] = 1
            self.boxes[self._get_box_id(row, col)][val] = 1
```

```
else:
        del self.rows[row][board[row][col]]
        del self.cols[col][board[row][col]]
        del self.boxes[self._get_box_id(row, col)][board[row][col]]
    board[row][col] = val
def _valid(self, board, row, col, val):
    return (val not in self.rows[row]) and \
           (val not in self.cols[col]) and \
           (val not in self.boxes[self._get_box_id(row, col)])
def _next_index(self, row, col):
    if col == self.n - 1:
       row += 1
        col = 0
    else:
        col += 1
   return row, col
def _dfs(self, board, row, col):
    if row >= self.n:
        self.success = True
        return
   new_row, new_col = self._next_index(row, col)
    if board[row][col] == 0:
        for val in range(1, self.n + 1):
            if self._valid(board, row, col, val):
                self._fill(board, row, col, val)
                self._dfs(board, new_row, new_col)
                if self.success:
                    return
                self._fill(board, row, col, 0)
    else:
        self._dfs(board, new_row, new_col)
def solveSudoku(self, board):
    for row in range(self.n):
        for col in range(self.n):
            if board[row][col] != 0:
                self._fill(board, row, col, board[row][col])
    self._dfs(board, 0, 0)
```

```
def max_val_board(board):
 row, col = len(board), len(board[1])
 val = [[0]*col for _ in range(row)]
 for r in range(row - 1, -1, -1):
    for c in range(col - 1, -1, -1):
     current = board[r][c]
     move_right = 0 if c == col - 1 else val[r][c + 1]
     move\_down = 0 if r == row - 1 else val[r + 1][c]
     val[r][c] = current + max(move_down, move_right)
     print(r, c, val[r][c])
```

return val[0][0]