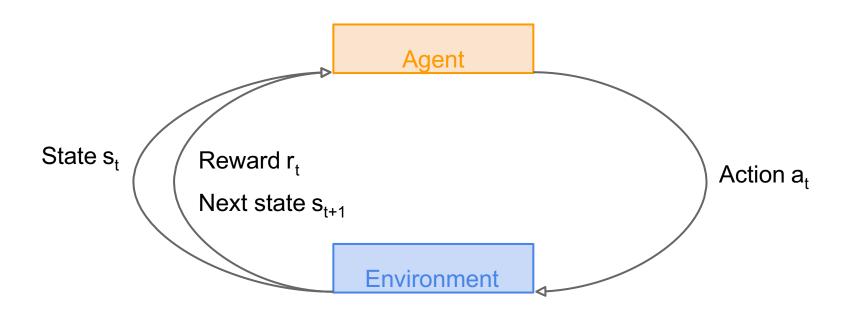
Artificial Intelligence Assignment 2

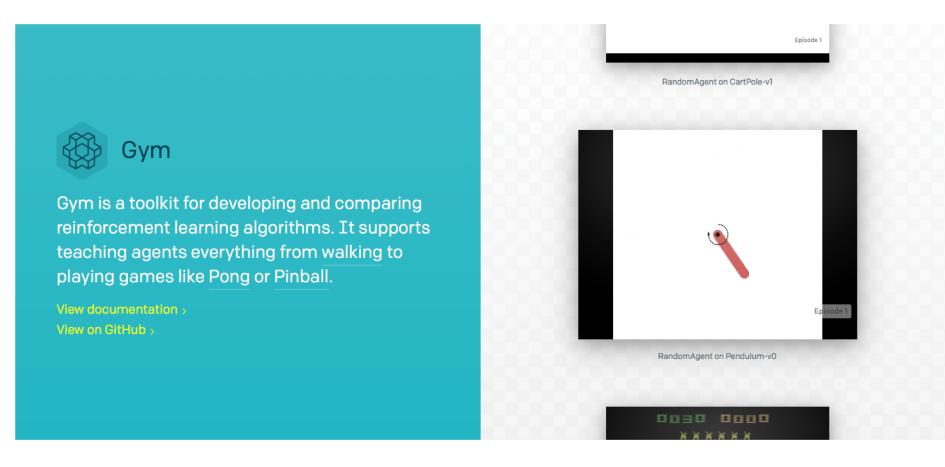
CSI4108-02 Spring, 2018

1. Introduction

 We will solve the toy text problem that is provided by the gym library using Reinforcement Learning.



2. OpenAl Gym





We provide the environment; you provide the algorithm. You can write your agent using your existing numerical computation library, such as TensorFlow or Theano.

2-1. Basic installation steps

Install gym library(pip install gym)

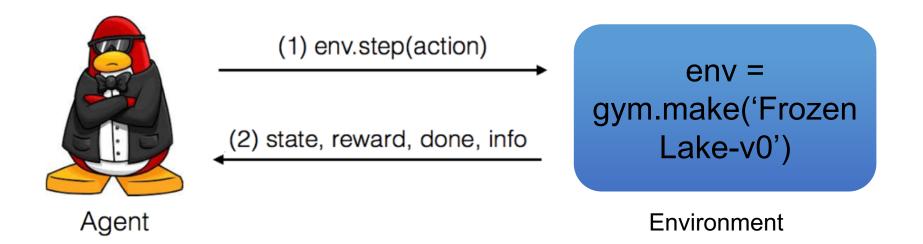
```
hansuhoui-MacBook-Pro:∼ hansuho$ pip3 install gym
Collecting gym
Requirement already satisfied: six in ./.pyenv/versions/3.5.4/lib/python3.5/site-packages (from gy
m) (1.11.0)
Requirement already satisfied: pyglet>=1.2.0 in ./.pyenv/versions/3.5.4/lib/python3.5/site-package
s (from gym) (1.3.2)
Requirement already satisfied: numpy>=1.10.4 in ./.pyenv/versions/3.5.4/lib/python3.5/site-package
s (from gym) (1.14.0)
Requirement already satisfied: requests>=2.0 in ./.pyenv/versions/3.5.4/lib/python3.5/site-package
s (from gym) (2.18.4)
Requirement already satisfied: future in ./.pyenv/versions/3.5.4/lib/python3.5/site-packages (from
pyglet = 1.2.0 - yym) (0.16.0)
Requirement already satisfied: idna<2.7,>=2.5 in ./.pyenv/versions/3.5.4/lib/python3.5/site-packag
es (from requests>=2.0->gym) (2.6)
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in ./.pyenv/versions/3.5.4/lib/python3.5/site
-packages (from requests>=2.0->gym) (3.0.4)
Requirement already satisfied: certifi>=2017.4.17 in ./.pyenv/versions/3.5.4/lib/python3.5/site-pa
ckages (from requests>=2.0->gym) (2018.4.16)
Requirement already satisfied: urllib3<1.23,>=1.21.1 in ./.pyenv/versions/3.5.4/lib/python3.5/site
-packages (from requests>=2.0->gym) (1.22)
Installing collected packages: gym
Successfully installed gym-0.10.5
```

Quick checking

```
hansuhoui-MacBook-Pro:~ hansuho$ python3
Python 3.5.4 (default, Jan 23 2018, 19:27:59)
[GCC 4.2.1 Compatible Apple LLVM 9.0.0 (clang-900.0.39.2)] on darwin
Type "help", "copyright", "credits" or "license" for more information.
>>> import gym
>>>
```

2-2. OpenAl Gym detail

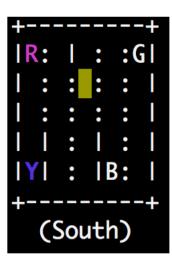
- Detail information about the gym library can be found below site.
- https://gym.openai.com/docs/



3. Detail

- In Assignment2, there are two problems.
 - FrozenLake8x8
 - Taxi

FrozenLake8x8



Taxi

- The agent controls the movement of a character in a grid world. Some tiles
 of the grid are walkable, and others lead to the agent falling into the water.
 Additionally, the movement direction of the agent is uncertain and only
 partially depends on the chosen direction. The agent is rewarded for
 finding a walkable path to a goal tile.
- Use Q-learning to solve this problem
- There are four actions (Left, Down, Right, Up).

```
SFFFFFFF
FFFFFFFF
FFFFFFFF
FFFHFFFF
FHHFFFHF
FHFFFFF
FHFFFFF
```

The episode ends when you reach the goal or fall in a hole. You receive a reward of 1 if you reach the goal, and zero otherwise.

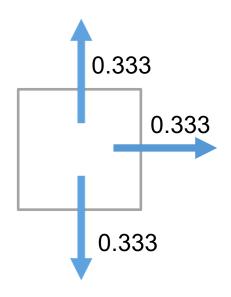
```
S : starting point, safe
F : frozen surface, safe
H : hole, fall to your doom
G : goal, where the frisbee is located
```

You need to use register to create environment.

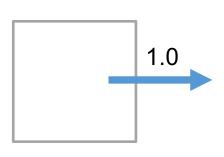
```
register(
    id='FrozenLake8x8-v3',
    entry_point='gym.envs.toy_text:FrozenLakeEnv',
    kwargs={'map_name': '8x8',
         'is_slippery':True}
)
env = gym.make('FrozenLake8x8-v3')
```

 Your code should work well when the 'is_slippery' is True and when it is False

 If the 'right' action is selected, the probability of the action depending on the 'is_slippery' is as follows.



Is_slippery: True



Is_slippery : False

Output

After learning Q-table, print the path to goal tile according to Q-table value on standard output.

You just print the path like <example>.

- Detail information about the FrozenLake8x8 can be found below site.
- https://gym.openai.com/envs/FrozenLake8x8-v0/

```
#### 1 action
  (Right)
FFFHFFFG
#### 2 action
  (Right)
FHFFHFHF
FFFHFFFG
#### 3 action
  (Right)
FFFFFHFF
FHFFHFHF
FFFHFFFG
```

3-1. Output Example

#### 1 action	#### 5 action	#### 9 action	#### 13 action	#### 1 action	#### 63 action	
(Down)	(Right)	(Down)	(Down)	(Right)	(Right)	
SFFFFFF	SFFFFFF	SFFFFFF	SFFFFFF	SFFFFFF	SFFFFFF	
FFFFFFF	FFFFFFFF	FFFFFFF	FFFFFFF	FFFFFFF	FFFFFFF	
FFFHFFFF	FFFHFFFF	FFFHFFFF	FFFHFFFF	FFFHFFF	FFFHFFFF	
FFFFFHFF	FFFFFHFF	FFFFFHFF		FFFFHFF	FFFFFHFF	
FFFHFFFF	FFFHFFFF	FFFHFFFF	FFFFHFF	FFFHFFFF	FFFHFFF	
FHHFFFHF	FHHFFFHF	FHHFFFHF	FFFHFFF	FHHFFFHF	FHHFFFHF	
FHFFHFHF	FHFFHFHF	FHFFHFHF	FHHFFFHF	FHFFHFHF	FHFFHFHF	
FFFHFFFG	FFFHFFFG	FFFHFFFG	FHFFHFH	FFFHFFFG	FFFHFFFG	
			FFFHFFFG			
#### 2 action	#### 6 action	#### 10 action		#### 2 action	#### 64 action	
(Right)	(Down)	(Right)	#### 14 action	(Up)	(Right)	
SFFFFFF	SFFFFFF	SFFFFFF	(Down)	SFFFFFF	SFFFFFF	
F <mark>F</mark> FFFFFF	FFFFFFF	FFFFFFF	SFFFFFF	FFFFFFF	FFFFFFF	
FFFHFFFF	FFFHEFFF	FFFHFFFF		FFFHFFFF	FFFHFFFF	
FFFFFHFF	FFFFFHFF	FFFFFHF	FFFFFFF	FFFFHFF	FFFFFHFF	
FFFHFFFF	FFFHFFFF	FFFHFFFF	FFFHFFFF	FFFHFFF	FFFHFFFF	
FHHFFFHF	FHHFFFHF	FHHFFFHF	FFFFHFF	FHHFFFHF	FHHFFFH F	
FHFFHFHF	FHFFHFHF	FHFFHFHF	FFFHFFF	FHFFHFHF	FHFFHFHF	
FFFHFFFG	FFFHFFFG	FFFHFFFG	FHHFFFHF	FFFHFFFG	FFFHFFFG	
			FHFFHFHF			
#### 3 action						
#### 3 accion	#### 7 action	#### 11 action	EEEHEEE6	#### 3 action	#### 65 action	
	#### 7 action	#### 11 action	FFFHFFF	#### 3 action	#### 65 action	
(Right) SFFFFFFF	(Right)	(Down)	FFFHFFF <mark>G</mark>	#### 3 action (Up)	(Right)	
(Right)	(Right) SFFFFFF	(Down) SFFFFFF	FFFHFFF <mark>G</mark>	#### 3 action (Up) SFFFFFFF	(Right) SFFFFFF	
(Right) SFFFFFFF	(Right) SFFFFFFF FFFFFFF	(Down) SFFFFFFF FFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF	(Right) SFFFFFFF FFFFFFF	dia alimanda Tura
(Right) SFFFFFFF FF <mark>F</mark> FFFFF	(Right) SFFFFFFF FFFFFFFF FFFHFFFFF	(Down) SFFFFFFF FFFFFFFF FFFHFFFF	<pre><'is_slippery': False></pre>	#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFHFFFF	<'is_slippery': True>
(Right) SFFFFFF FF <mark>F</mark> FFFF FFFHFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFF FFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFF FFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHHFFFHF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHHFFFHF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHHFFFHF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFF FFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFHF FHFFFHF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFHFFF FHHFFFHF FHFFFHFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFF FFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHHFFFHF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHHFFFHF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHHFFFHF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFF FFFFFFFFF FHHFFFFF FHFFFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFHFFFF FHFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFF FFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFF FFFFFF	(Down) SFFFFFFF FFFFFFFFF FFFFFFFFF FHHFFFFF FHFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFF FFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFF FFFFFFFFF FFFFFFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFFF FFFFFFFF FHFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHHFFFHF FHFFHF	(Right) SFFFFFFF FFFFFFFF FFFFFFFFF FFFFFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	(Down) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFF FFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFF FFFFFFFFFFFFFFFFFF	(Down) SFFFFFFF FFFFFFFFF FFFFFFFFFFFFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFHFFFF FHFFFFF FHFFFF FFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHFFFFF FHFFFFF FFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFF FFFFFFFFF FFFFFFFFF	(Down) SFFFFFFF FFFFFFFFF FFFFFFFFF FFFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FHHFFFFF FHFFFFG #### 4 action (Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFFFF	(Right) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Down) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF FFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Down) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFFF FFFFFFFFF FHHFFFFF FFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Down) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFF FFFFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	<'is_slippery': True>
(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	(Down) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF		#### 3 action (Up) SFFFFFFF FFFFFFFF FFFFFFFFF FFFFFFFFF FHHFFFFF FFFFFF	(Right) SFFFFFFF FFFFFFFFFFFFFFFFFFFFFFFFFFFF	<'is_slippery': True>

3-2. Taxi

- This task was introduced in [Dietterich2000] to illustrate some issues in hierarchical reinforcement learning. There are 4 locations (labeled by different letters) and your job is to pick up the passenger at one location and drop him off in another. You receive +20 points for a successful dropoff, and lose 1 point for every timestep it takes. There is also a 10 point penalty for illegal pick-up and drop-off actions.
- Use Q-learning to solve this problem
- There are six actions (West, South, East, North, Pickup, Dropoff).
- In this problem, You don't need to use register to create environment.

```
env = gym.make('Taxi-v2')
```

blue: passenger

magenta: destination

yellow: empty taxi

green: full taxi

other letters: locations

3-2. Taxi

Output

After learning Q-table, print the states until the end of the episode according to Q-table value on standard output.

You just print the path like <example>.

- Detail information about the Taxi can be found below site.
- https://gym.openai.com/envs/Taxi-v2/

```
## 1 action
(North)
(North)
### 3 action
(North)
(West)
```

3-2. Output Example

_	_	
#### 1 action	#### 5 action	#### 9 action
++	++	++
IR: : :G	IR: : :G	
1::::	1::::1	1::::1
1::::1	1::::1	
11:1:1	11:1:1	
Y : B:		Y : B:
++	++	++
(North)	(Pickup)	(North)
#### 2 action	#### 6 action	#### 10 actio
++	++	++
IR: : :G	IR: : :G	[[: : :G]
1_::::	1::::	!:::!
[: : : [·	<u> : : : ! </u>	! : : : !
1 1 : 1 : 1		: :
Y : B:	Y : B:	Y : B:
++	++	(Dropoff)
(West)	(North)	(broport)
#### 3 action	#### 7 action	
++	++	
IR: : :G	IR: : :G	
1::::1	1_: : : : 1	
1::::1	I <u>I</u> ::::!	
1 1 : 1 : 1	11:1:1	
Y : B:	Y : B:	
++		
(South)	(North)	
#### 4 action	#### 8 action	
++	+	
IR: 1 : :G1	IR: : :G	
1::::	! - : : : : !	
! : : : !		
	: : Y : B:	
¥ : B:	111 . ID:	
++	(North)	
(South)	(NOI CIT)	

3. Detail

Summary

Solve problems through reinforcement learning using the environment and functions given in the gym library.

We give you three Python files, then you have to write the code in that files.

- FrozenLake_true.py: The FrozenLake problem 'is_slippery' is ture
- FrozenLake_false.py: The FrozenLake problem 'is_slippery' is False
- Taxi.py: The Taxi problem

If you need another library(like numpy), you can use it.

3. Detail

Tips

- Use the Q-learning formula

$$sample = R(s, a, s') + \gamma \max_{a'} Q(s', a')$$

$$Q(s, a) \leftarrow (1 - \alpha)Q(s, a) + (\alpha) [sample]$$

- Set the noise appropriately for fast learning.
- Exploit & Explore

4. Submission

- Deliverables: 2013147xxx_1.zip
- Must include
 - FrozenLake_true.py
 - FrozenLake_false.py
 - Taxi.py
 - Other codes (If you necessary)

(Your code with detail comments)

5. Grading environment & Directions

- Language: Python
- We grade your score in Linux(Ubuntu 16.04)
- Python3 (>= 3.5.2)
- This is an individual project
- You should follow output format
- Never copy code
- You will get 0 points if you cheat
- If you do not use Reinforcement Learning, you will get 0 points

6. Grading policy

- FrozenLake_true.py
 - print a right path : 25pts
- FrozenLake_false.py
 - print a right path : 25pts
- Taxi.py
 - print right states : 50pts

7. Due Date

Due Date: 17/May/2018 23:59:00 KST

Delay Policy: -15pts per day

Pleae use YSCEC Q&A board to leave your question.