Homework 1 Solutions

This homework covers the following topics from class:

- Course Introduction
 - History, course methodology, compilers/interpreters/linkers
- Essential Python
 - Everything you (never learned but) need to know to write correct python code

PYTH1: Object Reference Semantics, Parameter Passing, Shallow Copying, Boxing (15 min)

Consider the following Python scripts:

Part A: (5 min) Here's our first script:

```
class Box:
    def __init__(self, value):
        self.value = value

def quintuple(num):
    num *= 5

def box_quintuple(box):
    box.value *= 5

num = 3
box = Box(3)

quintuple(num)
box_quintuple(box)
print(f"{num} {box.value}")
```

Running this script yields the output:

3 15

Explain, in detail, why box's value attribute was mutated but num was not. Your answer should explicitly mention Python's parameter passing semantics.

Hint: Check out the <u>Python documentation</u> to understand where class instances store their attributes.

Solution:

This answer is as instructive as we can make it - this level of detail will not be required on the exam.

In Python, everything is an object; every variable is essentially a pointer to an object. This is true for primitive data types too (including int, bool and str). Furthermore, Python uses pass by object reference for function parameters: if we have a function f with parameter x, and we invoke f by passing some variable a, then x merely points to whatever object that a points to. So, for example, when we refer to num *inside* the quintuple function, it is a separate pointer to the original object that the num *outside* the function points to. In other words, we have two different pointers that both point to the object 3.

Next, it is important to note that when you reassign a variable to a new value, you aren't changing the value of the old object pointed to by that variable - you make the variable point to the new object you're assigning it. In the case of the expression num *= 5 (which is syntactic sugar for num = num * 5), we change the num variable inside of the quintuple function such that it now points at the object 15. We do not modify the object that it previously pointed to (the int 3), nor do we change what the num variable outside of the function points to (which is why 3 is still printed).

However, custom classes in Python (like the Box class) do store their attributes separately (as alluded to in the hint, they're stored in a dictionary object named __dict__). So the expression box.value *= 5 modifies the value attribute on the same object pointed at by both box variables (i.e. the one inside the function and outside. Both variables point to the same object). This is why 15 is printed.

Part B: (10 min) Here's our second script, taken from a previous CS131 midterm exam:

```
class Comedian:
  def __init__(self, joke):
    self. joke = joke
  def change_joke(self, joke):
    self.__joke = joke
  def get joke(self):
    return self. joke
def process(c):
# Line A
 c[1] = Comedian("joke3")
 c.append(Comedian("joke4"))
 c = c + [Comedian("joke5")]
 c[0].change joke("joke6")
def main():
 c1 = Comedian("joke1")
 c2 = Comedian("joke2")
 com = [c1, c2]
 process(com)
 c1 = Comedian("joke7")
 for c in com:
  print(c.get_joke())
```

Part B.i: Assuming we run the main function, what will this program print out?

Answer for part i is below.

Part B.ii: Assuming we removed the comment on line A and replaced it with:

```
c = copy.copy(c)  # initiate a shallow copy
```

If we run the main function, what would the program print?

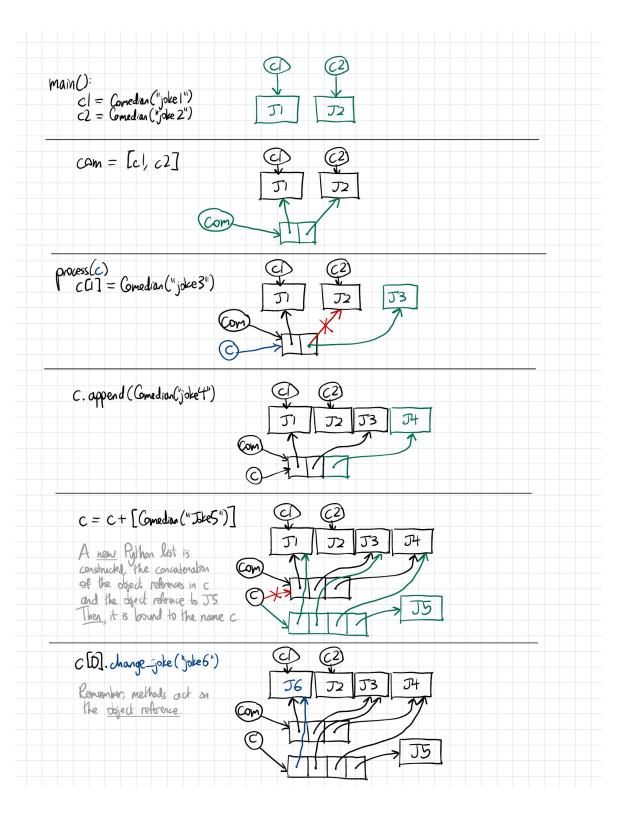
Answer for part ii is below.

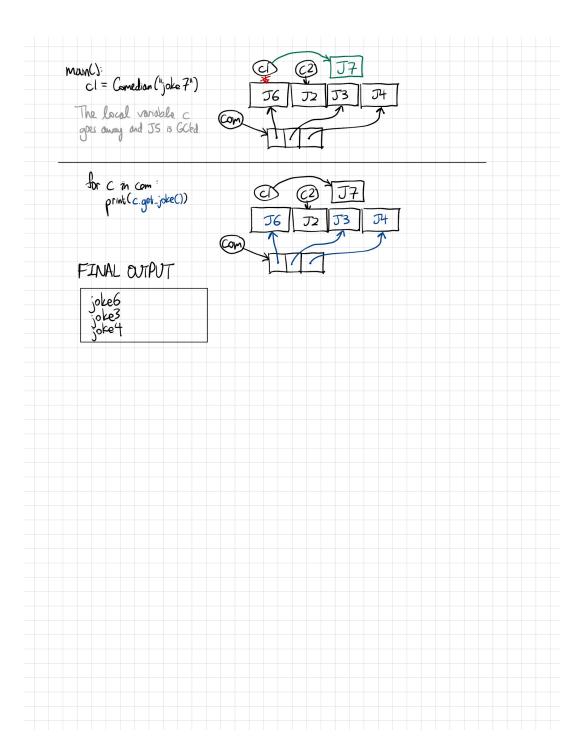
Answers for part i and ii, as written by our former student Vincent Lin:

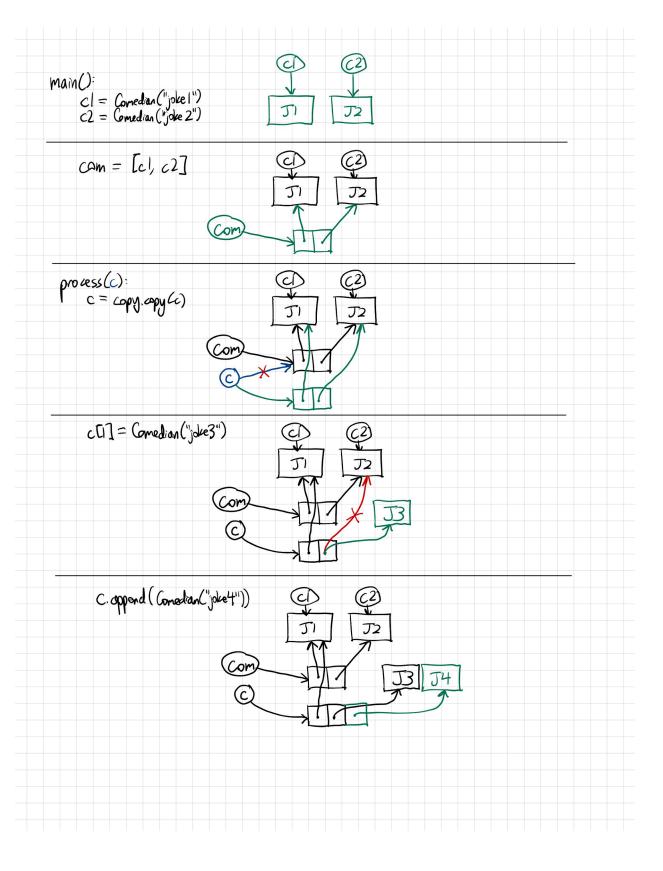
- Green describes a new allocation, extension, etc. as a result of the current line.
- Red describes something that has been removed as a result of the current line.
- Blue describes what's being referenced in a method call on the current line.

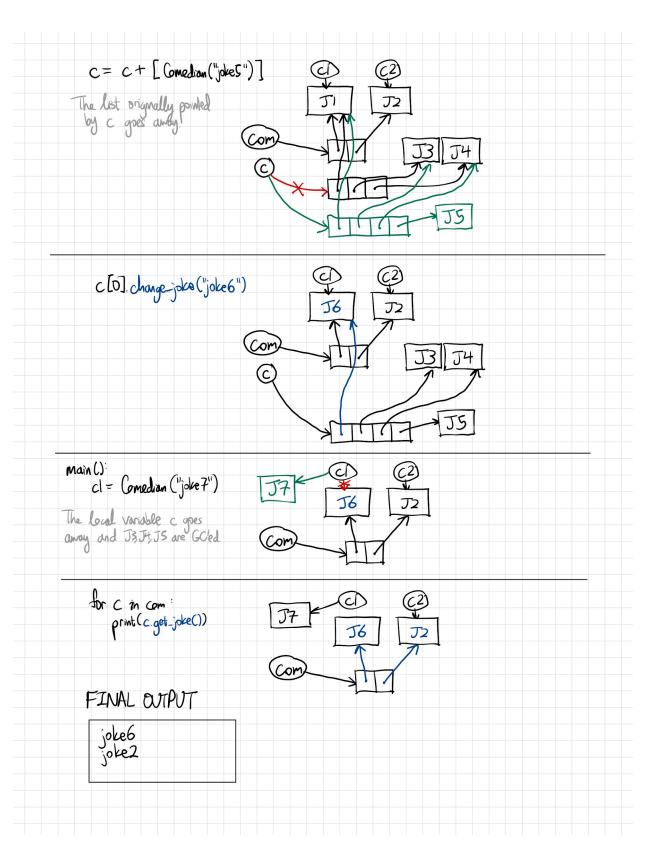
Also included are some notes in gray describing non-obvious behavior that Python performs behind the scenes.

Part (i) is below:









In General, for problems like these, there's a very simple trick: always, always draw pictures. You are NOT the interpreter. You are a human. Always draw pictures. Feel free to use these conventions:

- Boxes represent values, actual objects on the heap.
- Ellipses represent variables, names on the stack that are *bound* to the objects.

Drawing pictures effectively reduces the problem to following a bunch of arrows. This way, you'll *never* get lost.

The key part of this question is that it's testing object reference semantics. Remember that while Python made the syntax prettier, everything is really a pointer under the hood, including lists - a list is itself a pointer to an array of more pointers, which then point to the objects "stored" in the list.

Whenever you call a method like c.append, some_comedian.change_joke, etc. follow the arrow coming out of the ellipse with the variable name and perform the operation on the reference object(s). This works for more complex accesses like c[0].change_joke:

- 1. First follow the arrow out of c.
- 2. Choose the arrow out of array position o.
- 3. Follow that arrow and call change_joke on that object.

PYTH2: Duck Typing, Dunder Functions (5 min)

(5 min.) Consider the following output from the Python 3 REPL:

```
>>> class Foo:
...    def __len__(self):
...    return 10
...
>>> len(Foo())
10
>>> len("blah")
```

```
4
>>> len(5)
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
TypeError: object of type 'int' has no len()
```

Explain why Python allows you to pass an object of type Foo or str to len, but not one of type int. Your answer should explicitly reference Python's typing system.

Solution:

Python is duck-typed, which means that the type of an object is less important than the attributes and methods it defines. For example, functions in Python do not discriminate by requiring you to pass objects of a certain type - you can pass in anything you'd like. However, if you try to invoke a method or reference an attribute that does not exist on an object, Python will raise an error.

Built-in functions work the same way. In the example, you can see that for the len function to return without throwing an exception, the caller merely needs to pass in an object that has the __len__ function defined. In other words, the implementation of len("blah") at some point calls "blah".__len__(). int objects do not have a __len__() function, so the built-in len raises a TypeError.

PYTH3: Duck Typing, Easier To Ask For Forgiveness, Inheritance (9 min)

Consider the following Duck class, and two functions that check whether or not an object is a Duck:

```
class Duck:
    def __init__(self):
        pass # Empty initializer

def is_duck_a(duck):
    try:
        duck.quack()
        return True
    except:
        return False

def is_duck_b(duck):
```

Part A: (2 min.) Write a **new** class such that if we passed an instance of it to both functions, is_duck_a would return True but is_duck_b would return False.

return isinstance(duck, Duck)

Solution:

```
class RubberDuck:
   def quack(self):
    print("Squeak")
```

Part B: (2 min.) Write a **new** class such that if we passed an instance of it to both functions, is_duck_a would return False but is_duck_b would return True.

Solution:

class Mallard(Duck):
 pass # empty class, no new added methods/fields

Part C: (5 min.) Which function is more Pythonic: is_duck_a or is_duck_b? This reference may help provide some insight.

Solution:

is_duck_a is more Pythonic. In fact, both approaches have canonical names: "Easier to ask for forgiveness than permission" (EAFP) and "Look before you leap" (LBYL) respectively. Essentially, in Python it's considered more robust to make assumptions (e.g. that a Duck has a quack() method) and handle errors when those assumptions do not hold than to assert conditions ahead of time (am I dealing with a Duck object?) and carry on assuming they will always be true. For a concrete example of why, check out this old email from the Python mailing list.

PYTH4: Slicing (6 min)

Part A: (3 min.) Consider the following function that takes in a list of integers and another integer k and returns the largest sum you can obtain by summing k consecutive elements in the list. Fill in the blanks so the function works correctly.

Example:

```
largest_sum([3,5,6,2,3,4,5], 3) should return 14. largest_sum([10,-8,2,6,-1,2], 4) should return 10.
```

```
def largest_sum(nums, k):
```

```
if k < 0 or k > len(nums):
    raise ValueError
elif k == 0:
    return 0

max_sum = None
for i in range(len(nums)-k+1):
    sum = 0
    for num in _____:
        sum += num
    if _____:
        max_sum = sum
return max_sum
```

Solution:

```
def largest_sum(nums, k):
    if k < 0 or k > len(nums):
        raise ValueError
    elif k == 0:
        return 0

max_sum = None
    for i in range(len(nums)-k+1):
        sum = 0
        for num in nums[i:i+k]:
            sum += num
        if max_sum is None or sum > max_sum:
            max_sum = sum
        return max_sum
```

Part B: (3 min.) The function in part a) runs in O(nk) time where n is the length of nums, but we can use the **sliding window technique** to improve the runtime to O(n).

Imagine we know the sum of k consecutive elements starting at index i. We are interested in the next sum of k consecutive elements starting at index i+1. Notice that the only difference between these two sums is the element at index i (which becomes excluded) and the element at index i+k (which becomes included).

We can compute each next sum by subtracting the number that moved out of our sliding window and adding the one that moved in. Fill in the blanks so the function works correctly.

```
def largest_sum(nums, k):
    if k < 0 or k > len(nums):
        raise ValueError
    elif k == 0:
        return 0

sum = 0
for num in ______:
    sum += num

max_sum = sum
for i in range(0, len(nums)-k):
    sum -= _____
    sum += _____
    max_sum = max(sum, max_sum)
return max_sum
```

Solution:

```
def largest_sum(nums, k):
   if k < 0 or k > len(nums):
     raise ValueError
   elif k == 0:
     return 0
```

```
sum = 0
for num in nums[0:k]:
    sum += num

max_sum = sum
for i in range(0, len(nums)-k):
    sum -= nums[i]
    sum += nums[i+k]
    max_sum = max(sum, max_sum)
    return max_sum
```

PYTH5: Inheritance, Exceptions (9 min)

We're going to design an event-scheduling tool, like a calendar. Events have a start_time and an end_time. For simplicity, we will model points in time using ints that represent the amount of seconds past some reference time (normally, we would use the Unix epoch, which is January 1, 1970).

Part A: (3 min.) Design a Python class named Event which implements the above functionality. Include an initializer that takes in a start_time and end_time. If start_time >= end_time, raise a ValueError.

The following code:

```
event = Event(10, 20)
print(f"Start: {event.start_time}, End: {event.end_time}")

try:
   invalid_event = Event(20, 10)
   print("Success")
except ValueError:
```

```
print("Created an invalid event")
```

should print:

Start: 10, End: 20

Created an invalid event

Solution:

```
class Event:
    def __init__(self, start_time, end_time):
        if start_time >= end_time:
            raise ValueError
        self.start_time = start_time
        self.end_time = end_time
```

Part B: (3 min.) Write a Python class named Calendar that maintains a private list of scheduled events named __events.

It should:

- Include an initializer that takes in no parameters and initializes __events to an empty list.
- Have a method named get_events that returns the __events list.
- Have a method named add_event that takes in an argument event:
 - If the argument is not of type Event, do nothing and raise a TypeError.
 - Otherwise, add the event to the end of the events list.

The following code:

```
calendar = Calendar()
print(calendar.get_events())
calendar.add_event(Event(10, 20))
print(calendar.get_events()[0].start_time)
```

```
try:
   calendar.add_event("not an event")
except TypeError:
   print("Invalid event")
```

```
should print:
[]
10
Invalid event
```

Solution:

```
class Calendar:
    def __init__(self):
        self._events = []

    def get_events(self):
        return self._events

    def add_event(self, event):
        if not isinstance(event, Event):
            raise TypeError
        self._events.append(event)
```

Notice that we used isinstance even though, in the solution to question 6, we argued that it is more Pythonic to ask for forgiveness than to look before you leap. This would be the only way to check if an object is truly of a particular type. An alternative approach we could have taken would be to require that any object passed in implements the same set of methods as the Event class, rather than require only Event (or subclasses of Event) objects be passed in. This would enable duck typing to be used.

Part C: (3 min.) Consider the following subclass of Calendar:

```
class AdventCalendar(Calendar):
    def __init__(self, year):
        self.year = year

advent_calendar = AdventCalendar(2022)
print(advent_calendar.get_events())
```

Running this code as-is (assuming you have a correct Calendar implementation) yields the following output:

```
AttributeError: 'AdventCalendar' object has no attribute '_events'. Did you mean: 'get_events'?
```

Explain why this happens, and list two different ways you could fix the code **within the class** so the snippet instead prints:

[]

Solution:

While the AdventCalendar class does inherit from the Calendar class, it does not explicitly invoke its base class' initializer via super().__init__. Because of this, Calendar's initializer (which adds an empty events list to the instance) is never run. Notice, however, that we do still inherit the methods defined on Calendar (namely get_events).

To fix this, inside AdventCalendar's __init__ function we can either add an explicit call to the super initializer:

```
super().__init__()
```

or we can just add an events attribute ourselves. Of course, calling the superclass initializer would be better practice (since it leads to less repetitive code, especially in cases where the base class initializer has additional logic).

self._events = []

PYTH6: Interpreted vs Compiled Languages (6 min)

Suppose we have 2 languages. The first is called I-Lang and is an interpreted language. The interpreter receives lines of I-Lang and efficiently executes them line by line. The second is called C-Lang and is a compiled language. Assume that it takes the same time to write an I-Lang script as a C-Lang program if both perform the same function.

Part A: (2 min.) Suppose we have an I-Lang script and a C-Lang executable that functionally perform the exact same thing. If we execute both at the same time, which do you expect to be faster and why?

Solution:

We would expect the C-Lang executable to be faster. The C-Lang program is compiled directly to machine code, whereas the interpreted I-Lang code is translated into lower-level representations at run-time. This extra step creates additional overhead which will inevitably cause it to be slower than the pre-compiled code.

Part B: (2 min.) Suppose Jamie and Tim are two equally competent students developing a web server that sends back a simple, plaintext HTML page. Jamie writes her server in I-Lang, whereas Tim writes his in C-Lang. Assuming that the server will be deployed locally, who will most likely have the server running first, and why?

Solution:

We would expect Jamie to get the web server running first. From start to finish, the only difference between both development processes is that Jamie needs to execute her code using the interpreter, whereas Tim needs to compile his into an executable. In general, the compilation process takes a nontrivial amount of time, whereas the I-Lang interpreter can begin executing lines of code right away.

It could also be correct to argue that Tim would be able to get the server running first. If the I-Lang interpreter was extremely slow, or if the C-Lang compiler was abnormally fast, then as in part a), the additional overhead associated with interpreting each line of I-Lang may be enough to make Tim the winner. For most modern languages, however, compilation (which involves preprocessing, compiling, assembling, and linking) is a process that takes much longer than simply interpreting lines of code.

Part C: (2 min.) Jamie and Tim have a socialite friend named Connie who uses a fancy new SmackBook Pro. The SmackBook Pro has a special kind of chip called the N1, which has a proprietary machine language instruction set (<u>ISA</u>) completely unique to anything else out there. If you are familiar with Rosetta, assume the SmackBook Pro has **no** built-in emulator/app translator. Jamie and Tim have less-fancy computers with Intel chips.

Connie has a native copy of the I-Lang interpreter on her computer. Jamie sends Connie her web server script, and Tim sends Connie his pre-compiled executable. Will Connie be able to execute Jamie's script? What about Tim's executable?

Solution:

Connie will be able to execute Jamie's script because she already has a working version of the I-Lang interpreter on her computer. However, she will not be able to run Tim's executable, which has already been compiled into machine code that is tailored for Intel's ISA - it isn't portable.

PYTH7: Numpy (10 min)

(10 min.) Consider the following code snippet that compares the performance of <u>matrix</u> <u>multiplication</u> using a hand-coded Python implementation versus using <u>numpy</u>:

```
import numpy as np
import time
from random import randint
def dot_product(a, b):
  result = 0
 for a i, b i in zip(a, b):
    result += a i * b i
  return result
def matrix multiply(matrix, vector):
  return [dot product(row vector, vector) for row vector in matrix]
# Generate a 1000 element vector, and a 1000 x 1000 element matrix
vector = [randint(0, 10) for in range(1000)]
matrix = [[randint(0, 10) for in range(1000)] for in range(1000)]
# Create numpy arrays
np_vector = np.array(vector)
np matrix = np.array(matrix)
# Multiply the matrix and vector (using our hand-coded implementation)
start = time.time()
matrix multiply(matrix, vector)
```

```
end = time.time()

# Multiply the matrix and vector (using numpy)

np_start = time.time()

np_matrix.dot(np_vector)

np_end = time.time()

print(f"Hand-coded implementation took {end - start} seconds")

print(f"numpy took {np_end - np_start} seconds")
```

If we run this code, the numpy operation is invariably about 100 times faster than the call to matrix_multiply:

Hand-coded implementation took 0.043227195739746094 seconds numpy took 0.0004067420959472656 seconds

Assuming that the implementations of dot_product and matrix_multiply are reasonably optimized, provide an explanation for this discrepancy in performance.

Hint: Take a look at <u>this page</u> from the numpy docs to understand a bit more about how the package is implemented.

Solution:

The key to answering this question is to recognize that under the hood (as referenced by the hint), numpy's implementation (which includes functions like dot) heavily relies on pre-compiled, optimized C code. There are therefore two main reasons (only the first of which we would expect you to reasonably elaborate on on an exam) that numpy is significantly faster.

Firstly, as discussed in previous questions, compiled code is much more performant than interpreted code, because the latter involves the overhead of translation to lower-level representations at runtime. This is especially pertinent for our code, which will call the dot_product function at least 1000 times. For each invocation, the Python interpreter will convert the bytecode corresponding to dot_product into machine-executable instructions. This process has already been done ahead of time for the C code.

Secondly, the underlying memory representation for numpy arrays is much more efficient than that of Python lists. Recall that in Python, everything is an object, which means that the matrix variable actually contains one million pointers. Not only is this less space efficient, there is the chance of poor <u>spatial locality</u>: the objects pointed to in the matrix might be scattered throughout RAM. However, numpy provides support for contiguous arrays where each element is actually next to each other in memory (as in C). This enables more efficient caching techniques at the CPU level and chip-specific optimizations such as <u>SIMD</u>.

PYTH8: Class vs Instance Variables (5 min)

(5 min.) Consider the following code snippet that uses both class variables and instance variables:

```
class Joker:
  joke = "I dressed as a UDP packet at the party. Nobody got it."

def change_joke(self):
  print(f'self.joke = {self.joke}')
  print(f'Joker.joke = {Joker.joke}')
  Joker.joke = "How does an OOP coder get wealthy? Inheritance."
  self.joke = "Why do Java coders wear glasses? They can't C#."
  print(f'self.joke = {self.joke}')
```

```
print(f'Joker.joke = {Joker.joke}')

j = Joker()
print(f'j.joke = {j.joke}')
print(f'Joker.joke = {Joker.joke}')
j.change_joke()
print(f'j.joke = {j.joke}')
print(f'j.joke = {Joker.joke}')
```

Part A: (2 min) What do you expect this program to print out?

Solution:

```
#note: newlines added for clarity
j.joke = I dressed as a UDP packet at the party. Nobody got it.

Joker.joke = I dressed as a UDP packet at the party. Nobody got it.

self.joke = I dressed as a UDP packet at the party. Nobody got it.

Joker.joke = I dressed as a UDP packet at the party. Nobody got it.

self.joke = Why do Java coders wear glasses? They can't C#.

Joker.joke = How does an OOP coder get wealthy? Inheritance.

j.joke = Why do Java coders wear glasses? They can't C#.

Joker.joke = How does an OOP coder get wealthy? Inheritance.
```

Part B: (3 min) What does each print statement actually print out? Why?

Solution:

The correct answer is above, so this part will just be an explanation of the above.

Note that modifying a class variable on the class namespace, modifies all the instances of the class.

Also, for the field self. joke, if there is no instance variable joke in the self instance, then self. joke will reference the class variable joke on the class that self is an instance of (in this case, Joker). However, if an instance variable self. joke exists, it will shadow the class variable of the same name, overriding and hiding it.

The line:

self.joke = "Why do Java coders wear glasses? They can't C#." creates a new joke variable that gets added to the instance of the Joker class.

So, the first 4 lines are the same, since self. joke and Joker. joke refer to the same class variable.

Then, line 5 refers to the (newly created!) instance variable, and line 6 refers to the changed class variable.

Finally, line 7 refers to the instance variable and line 8 refers to the class variable, so the same situation as lines 5 and 6.