
TALLER 2

Grafos, Complejidad Computacional, Programación Dinámica

Laura Valentina Castaño Morales- 1010229518

Ing. Germán Hernández

(Docente)

UNIVERSIDAD NACIONAL DE COLOMBIA

SEDE BOGOTÁ

FACULTAD DE INGENIERÍA

DEPARTAMENTO DE SISTEMAS E INDUSTRIAL

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Desarrollo

1. Considerando el grafo de la Figura 1

a. Algoritmo de Dijkstra

1. Inicializamos todas las distancias en D con un valor infinito debido a que son desconocidas al principio, la del nodo start se debe colocar en 0 debido a que la distancia de start a start sería 0.
2. Sea $a = \text{start}$ (tomamos a como nodo actual).
3. Visitamos todos los nodos adyacentes de a, excepto los nodos marcados, llamaremos a estos nodos no marcados v_i .
4. Para el nodo actual, calculamos la distancia desde dicho nodo a sus vecinos con la siguiente fórmula: $dt(v_i) = D_a + d(a, v_i)$. Es decir, la distancia del nodo ' v_i ' es la distancia que actualmente tiene el nodo en el vector D más la distancia desde dicho el nodo ' a ' (el actual) al nodo v_i . Si la distancia es menor que la distancia almacenada en el vector, actualizamos el vector con esta distancia tentativa. Es decir: $\text{newDuv} = D[u] + G[u][v]$

```
if newDuv < D[v]:
```

```
    P[v] = u
```

```
    D[v] = newDuv
```

```
    updateheap(Q, D[v], v)
```

5. Marcamos como completo el nodo a.
6. Tomamos como próximo nodo actual el de menor valor en D (lo hacemos almacenando los valores en una cola de prioridad) y volvemos al paso 3 mientras existan nodos no marcados.

Una vez terminado al algoritmo, D estará completamente lleno.

```
{1: 0, 2: 12, 3: 8, 4: 10, 5: 14, 6: 10, 7: 18, 8: 14, 9: 13, 10: 15}
{1: inf, 2: 0, 3: 8, 4: 17, 5: 6, 6: 8, 7: 10, 8: 12, 9: 11, 10: 11}
{1: inf, 2: 4, 3: 0, 4: 11, 5: 6, 6: 2, 7: 10, 8: 6, 9: 5, 10: 7}
{1: inf, 2: 11, 3: 7, 4: 0, 5: 5, 6: 1, 7: 9, 8: 5, 9: 4, 10: 6}
{1: inf, 2: 6, 3: 2, 4: 11, 5: 0, 6: 2, 7: 4, 8: 6, 9: 5, 10: 5}
{1: inf, 2: 10, 3: 6, 4: 9, 5: 4, 6: 0, 7: 8, 8: 4, 9: 3, 10: 5}
{1: inf, 2: 26, 3: 22, 4: 31, 5: 20, 6: 22, 7: 0, 8: 20, 9: 25, 10: 1}
{1: inf, 2: 6, 3: 2, 4: 11, 5: 0, 6: 2, 7: 4, 8: 0, 9: 5, 10: 5}
{1: inf, 2: 16, 3: 12, 4: 6, 5: 10, 6: 7, 7: 14, 8: 10, 9: 0, 10: 2}
{1: inf, 2: inf, 3: inf, 4: inf, 5: inf, 6: inf, 7: inf, 8: inf, 9: inf, 10: 0}
```

b. Algoritmo de Bellman-Ford

1. Inicializamos el grafo. Ponemos distancias a INFINITO en todos los nodos menos en el nodo origen que tiene distancia 0.
2. Tenemos un diccionario de distancias finales y un diccionario de padres.

3. Visitamos cada arista del grafo tantas veces como número de nodos -1 haya en el grafo
4. Comprobamos si hay ciclos negativos.

La salida es una lista de los vértices en orden de la ruta más corta

```
{1: 0, 2: 12, 3: 8, 4: 10, 5: 14, 6: 10, 7: 18, 8: 14, 9: 13, 10: 15}
{1: inf, 2: 0, 3: 8, 4: 17, 5: 6, 6: 8, 7: 10, 8: 12, 9: 11, 10: 11}
{1: inf, 2: 4, 3: 0, 4: 11, 5: 6, 6: 2, 7: 10, 8: 6, 9: 5, 10: 7}
{1: inf, 2: 11, 3: 7, 4: 0, 5: 5, 6: 1, 7: 9, 8: 5, 9: 4, 10: 6}
{1: inf, 2: 6, 3: 2, 4: 11, 5: 0, 6: 2, 7: 4, 8: 6, 9: 5, 10: 5}
{1: inf, 2: 10, 3: 6, 4: 9, 5: 4, 6: 0, 7: 8, 8: 4, 9: 3, 10: 5}
{1: inf, 2: 26, 3: 22, 4: 31, 5: 20, 6: 22, 7: 0, 8: 20, 9: 25, 10: 1}
{1: inf, 2: 6, 3: 2, 4: 11, 5: 0, 6: 2, 7: 4, 8: 0, 9: 5, 10: 5}
{1: inf, 2: 16, 3: 12, 4: 6, 5: 10, 6: 7, 7: 14, 8: 10, 9: 0, 10: 2}
{1: inf, 2: inf, 3: inf, 4: inf, 5: inf, 6: inf, 7: inf, 8: inf, 9: inf, 10: 0}
```

c. Algoritmo de Floyd-Warshall

El algoritmo de Floyd-Warshall compara todos los posibles caminos a través del grafo entre cada par de vértices.

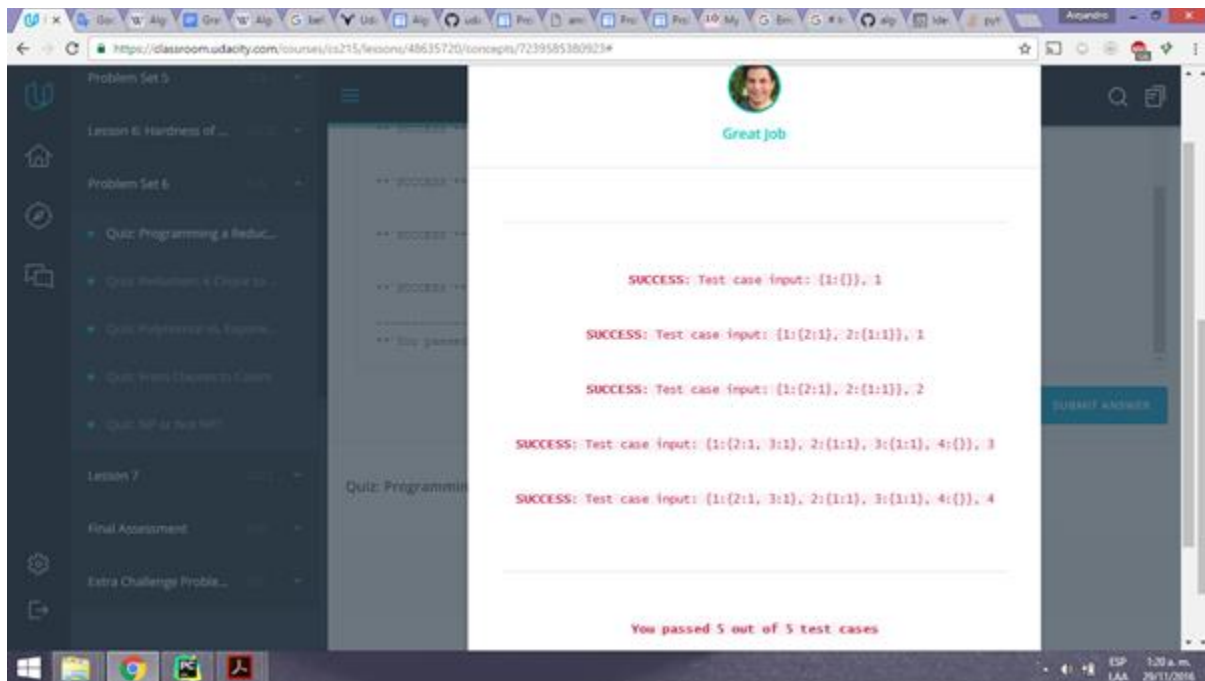
1. Formar las matrices iniciales C y D.
2. Se toma $k=1$.
3. Se selecciona la fila y la columna k de la matriz C y entonces, para i y j, con $i \neq k$, $j \neq k$ e $i \neq j$, hacemos:
4. Si $(C_{ik} + C_{kj}) < C_{ij} \rightarrow D_{ij} = D_{kj}$ y $C_{ij} = C_{ik} + C_{kj}$
5. En caso contrario, dejamos las matrices como están.
6. Si $k \leq n$, aumentamos k en una unidad y repetimos el paso anterior, en caso contrario para las iteraciones.
7. La matriz final C contiene los costes óptimos para ir de un vértice a otro, mientras que la matriz D contiene los penúltimos vértices de los caminos óptimos que unen dos vértices, lo cual permite reconstruir cualquier camino óptimo para ir de un vértice a otro.

```

1 10 3 2 5 4 7 6 9 8
('1', [0, 15, 8, 12, 14, 10, 18, 10, 13, 14])
('10', [inf, 0, inf, inf, inf, inf, inf, inf, inf, inf])
('3', [inf, 7, 0, 4, 6, 11, 10, 2, 5, 6])
('2', [inf, 11, 8, 0, 6, 17, 10, 8, 11, 12])
('5', [inf, 5, 2, 6, 0, 11, 4, 2, 5, 6])
('4', [inf, 6, 7, 11, 5, 0, 9, 1, 4, 5])
('7', [inf, 1, 22, 26, 20, 31, 0, 22, 25, 20])
('6', [inf, 5, 6, 10, 4, 9, 8, 0, 3, 4])
('9', [inf, 2, 12, 16, 10, 6, 14, 7, 0, 10])
('8', [inf, 5, 2, 6, 0, 11, 4, 2, 5, 0])

```

2.



```
# This function should use the k_clique_decision function
```

```
# to solve the independent set decision problem
```

```
def independent_set_decision(H, s):
```

```
    # your code here
```

```
    G = {}
```

```

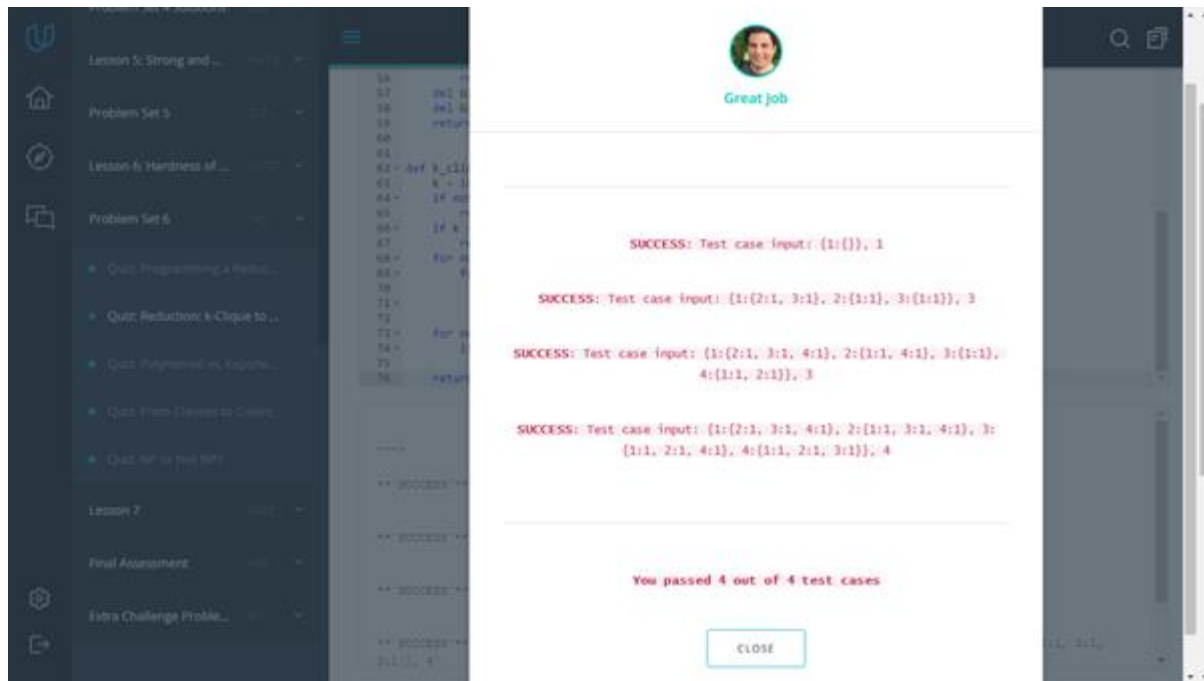
all_nodes = H.keys()
for v in H.keys():
    G[v] = {}
    for other in list(set(all_nodes) - set(H[v].keys()) - set([v])):
        G[v][other] = 1
print G
return k_clique_decision(G, s)

def test():
    H={}
    edges = [(1,2), (1,4), (1,7), (2,3), (2,5), (3,5), (3,6), (5,6), (6,7)]
    for u,v in edges:
        make_link(H,u,v)
    for i in range(1,8):
        print(i, independent_set_decision(H, i))

test()

```

2. Reduction: k-Clique to Decision



```
def k_clique(G, k):
    k = int(k)

    if not k_clique_decision(G, k):
        #your code here
        return False

    if k == 1:
        return [G.keys()[0]]

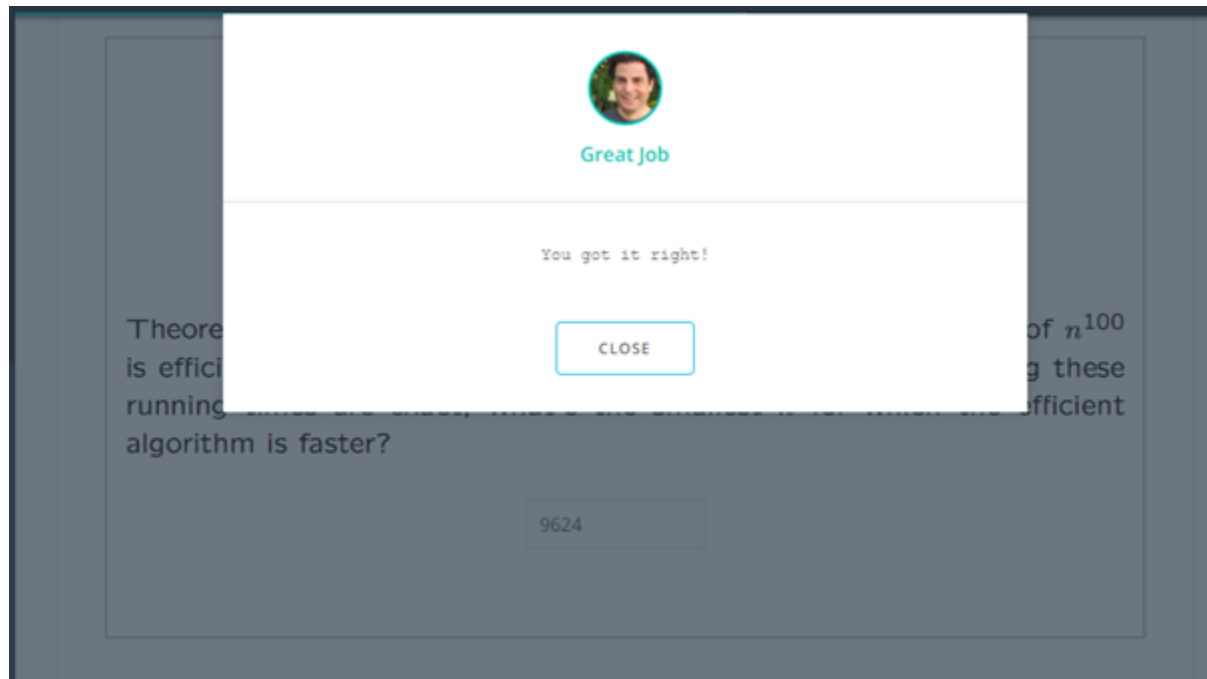
    for node1 in G.keys():
        for node2 in G[node1].keys():
            G = break_link(G, node1, node2)

            if not k_clique_decision(G, k):
                G = make_link(G, node1, node2)

    for node in G.keys():
        if len(G[node]) == 0:
            del G[node]

    return G.keys()
```

3. Polynomial vs. Exponential



4. From Clauses to Colors

From Clauses to Colors

In the reduction from 3-SAT to 3-COLORABILITY, we talked about a way of converting a 3-SAT problem with x variables and y clauses into a graph with n nodes and m edges. Give a formula for n and m . (Fill in the boxes to complete the equation. See the example given below.)

$$\begin{array}{rclclclcl} n & = & \boxed{2} & x & + & \boxed{6} & y & + & \boxed{3} \\ m & = & \boxed{3} & x & + & \boxed{12} & y & + & \boxed{3} \end{array}$$

(ex. $n = 4x + 10y + 8$)

5. NP or Not NP?

NP or Not NP? That is the Question

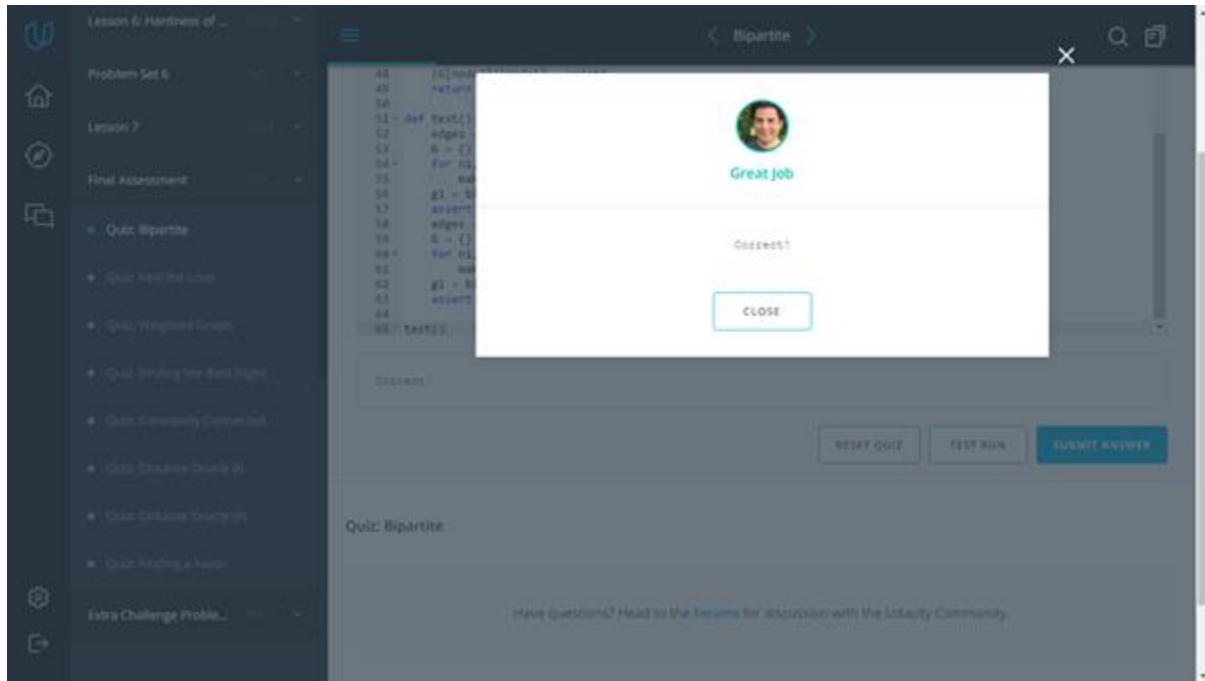
Select all the problems below that are in NP. Hint: Think about whether or not each one has a short accepting certificate.

- ☐ **Connectivity:** Is there a path from x to y in G ?
- ☐ **Short path:** Is there a path from x to y in G that is no more than k steps long?
- ☒ **Fewest colors:** Is k the absolute minimum number of colors with which G can be colored?
- ☒ **Near Clique:** Is there a group of k nodes in G that has at least s pairs that are connected?
- ☐ **Partitioning:** Can we group the nodes of G into two groups of size $n/2$ so that there are no more than k edges between the two groups.
- ☒ **Exact coloring count:** Are there exactly s ways to color graph G with k colors?

SUBMIT ANSWER

3. dsa

1. Bipartite



```
from collections import deque

def bipartite(G):
    # your code here
    # return a set

    if not G:
        return None

    start = next(G.iterkeys())
    lfrontier, rexplored, L, R = deque([start]), set(), set(), set()

    while lfrontier:
        head = lfrontier.popleft()

        if head in rexplored:
            return None
```

```

    if head in L:
        continue

    L.add(head)

    for successor in G[head]:
        if successor in reexplored:
            continue

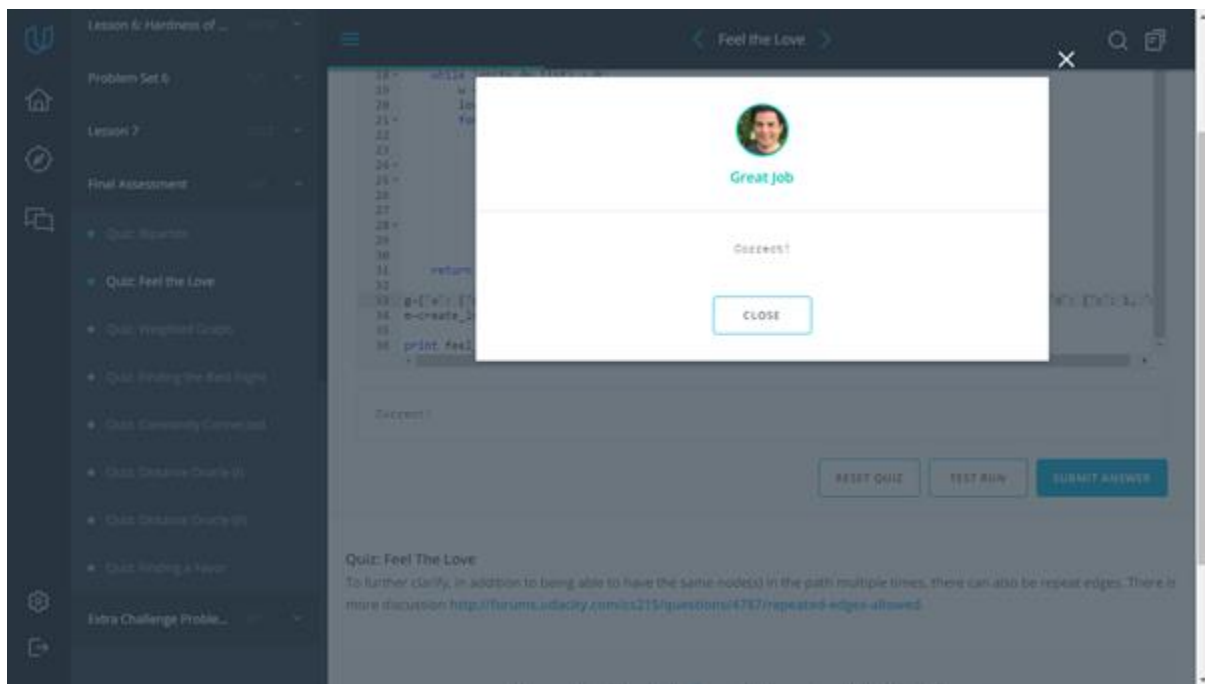
        R.add(successor)
        reexplored.add(successor)

        for nxt in G[successor]:
            lfrontier.append(nxt)

    return L

```

2. Feel the Love



```

def feel_the_love(G, i, j):

```

```

# return a path (a list of nodes) between `i` and `j`,
# with `i` as the first node and `j` as the last node,
# or None if no path exists

result = create_love_paths(G, i)

if j in result:
    return result[j][1]

else:
    return None


def create_love_paths(G, v):
    love_so_far = {}
    love_so_far[v] = (0, [v])
    to_do_list = [v]
    while len(to_do_list) > 0:
        w = to_do_list.pop(0)
        love, path = love_so_far[w]
        for x in G[w]:
            new_path = path + [x]
            new_love = max([love, G[w][x]])
            if x in love_so_far:
                if new_love > love_so_far[x][0]:
                    love_so_far[x] = (new_love, new_path)
                    if x not in to_do_list: to_do_list.append(x)
            else:
                love_so_far[x] = (new_love, new_path)
                if x not in to_do_list: to_do_list.append(x)
    return love_so_far

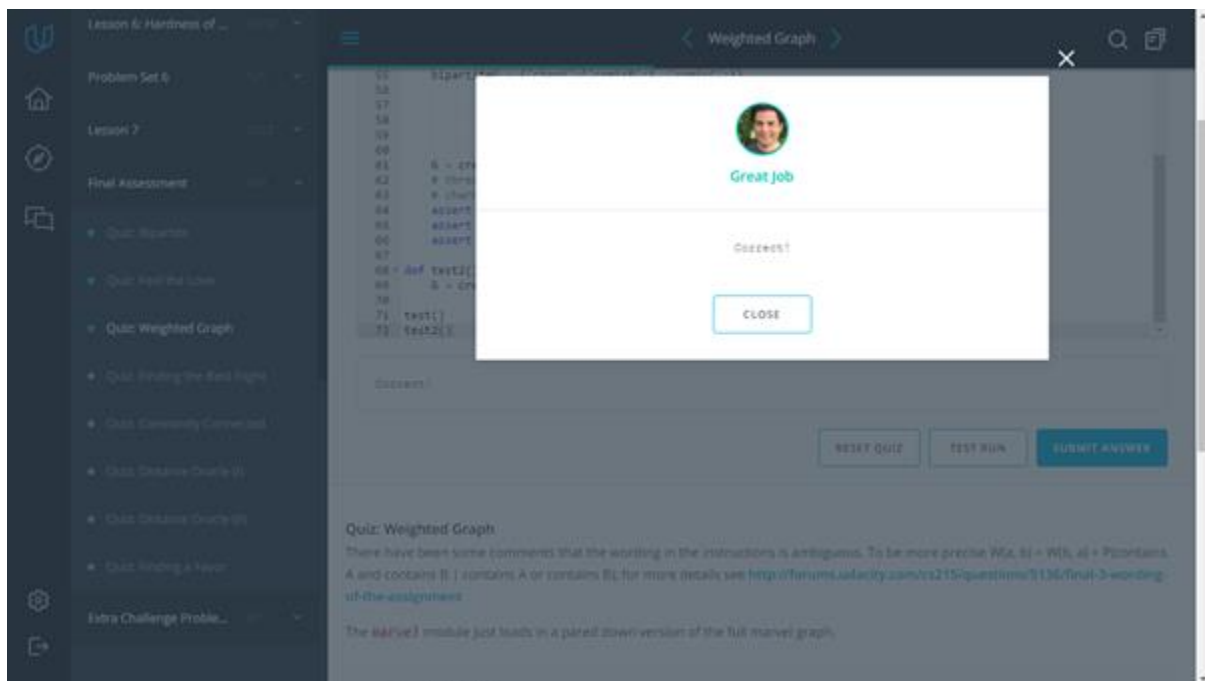
```

```
g={'a': {'c': 1}, 'c': {'a': 1, 'b': 1, 'e': 1, 'd': 1}, 'b': {'c': 1},
  'e': {'c': 1, 'd': 2}, 'd': {'c': 1, 'e': 2}}
```

```
m=create_love_paths(g, 'a')
```

```
print feel_the_love(g, 'a', 'e')
```

3. Weighted Graph



```
def create_weighted_graph(bipartiteG, characters):
    comic_size = len(set(bipartiteG.keys()) - set(characters))
    # your code here

    AB = {}

    for ch1 in characters:
        if ch1 not in AB:
            AB[ch1] = {}

        for book in bipartiteG[ch1]:
```

```

        for ch2 in bipartiteG[book]:
            if ch1 != ch2:
                if ch2 not in AB[ch1]:
                    AB[ch1][ch2] = 1
                else:
                    AB[ch1][ch2] += 1

contains = {}

for ch1 in characters:
    if ch1 not in contains:
        contains[ch1] = {}
        contains[ch1] = len(bipartiteG[ch1].keys())

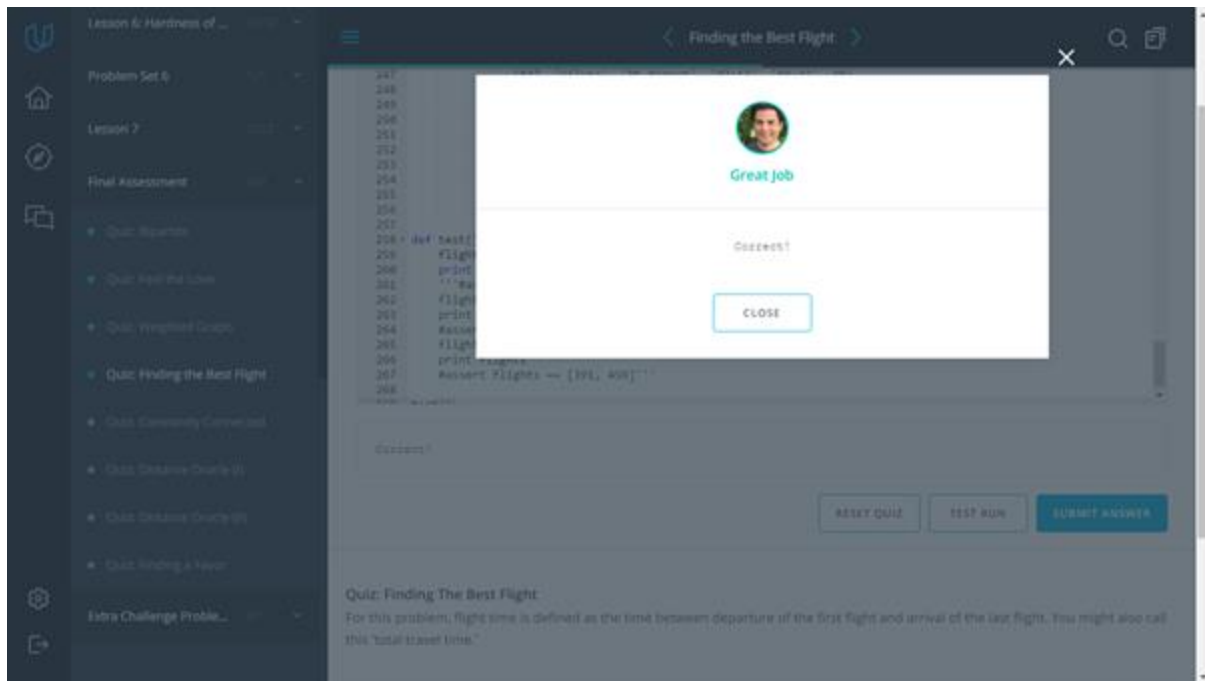
G = {}

for ch1 in characters:
    if ch1 not in G:
        G[ch1] = {}
        for book in bipartiteG[ch1]:
            for ch2 in bipartiteG[book]:
                if ch2 != ch1:
                    G[ch1][ch2] = (0.0 + AB[ch1][ch2]) / (contains[ch1] +
contains[ch2] - AB[ch1][ch2])

    return G

```

4. Finding the best Flight



```
import heapq
```

```
def find_best_flights(flights, origin, destination):
```

```
    G = make_graph(flights)
```

```
    R = find_route(G, origin, destination)
```

```
    return R
```

```
def make_graph(flights):
```

```
    edges = {}
```

```
    for (flight_number, origin, dest, take_off, landing, cost) in flights:
```

```
        to = make_time(take_off)
```

```
        land = make_time(landing)
```

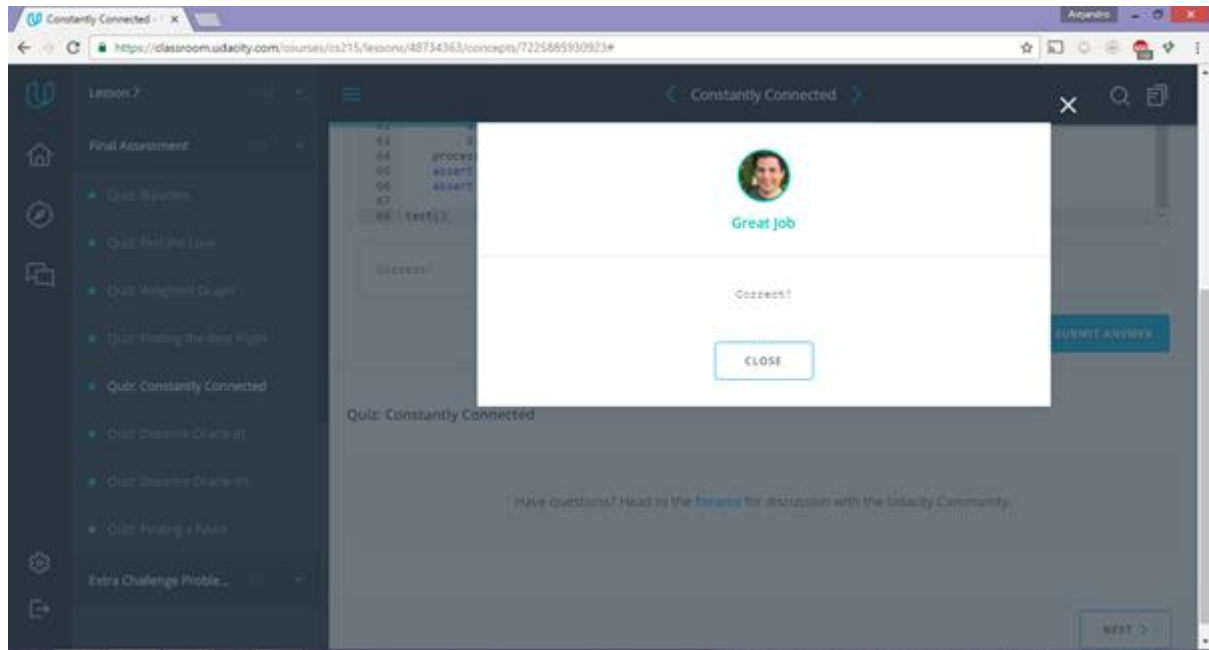
```
        edges[flight_number] = {'origin':origin, 'dest':dest,  
'take_off':to, 'land':land, 'cost':cost}
```

```
        if origin not in edges:
```



```
return None
```

5. Constantly Connected



```
conns = {}
```

```
def process_graph(G):  
    # your code here  
    global conns  
    conns = {}  
    groupId = 0  
    nodes = G.keys()  
    while len(conns) < len(G):  
        c_node = nodes.pop()  
        if c_node not in conns: conns[c_node] = groupId
```



```

open_list = [c_node]

while open_list:
    reached = open_list.pop()

    for neighbor in G[reached]:
        if neighbor not in conns:
            open_list.append(neighbor)
            conns[neighbor] = groupId

            if neighbor in nodes:
                del nodes[nodes.index(neighbor)]

groupId += 1

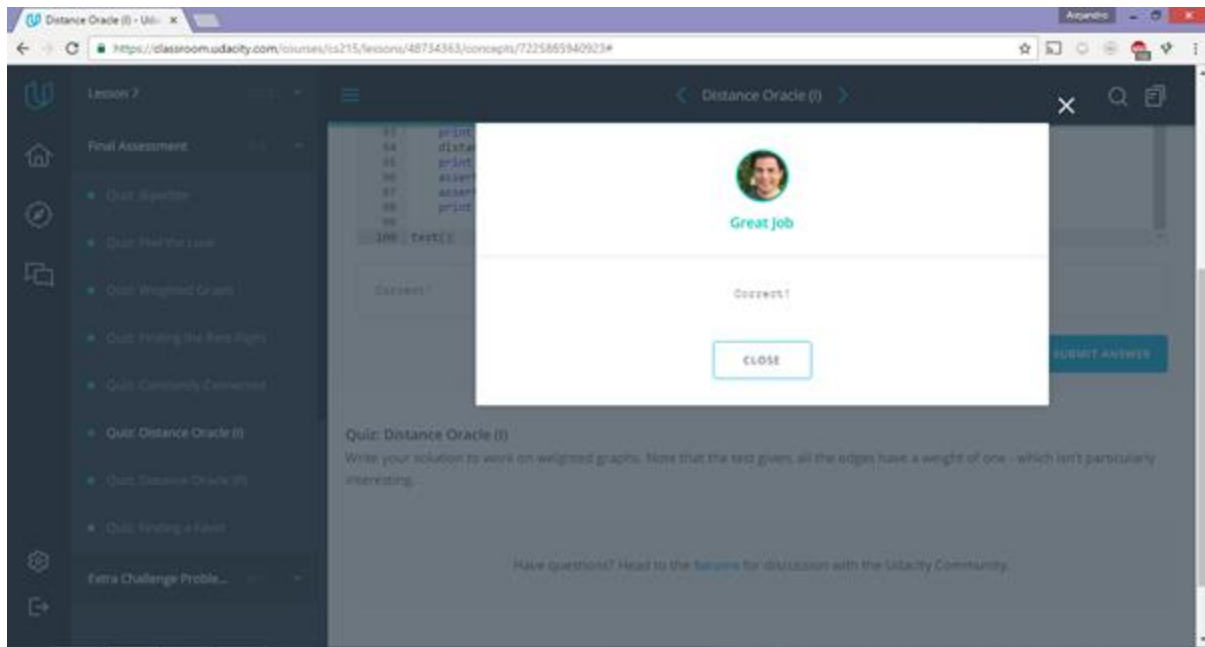
#
# When being graded, `is_connected` will be called
# many times so this routine needs to be quick
#
def is_connected(i, j):
    # your code here

    global conns

    return conns[i] == conns[j]

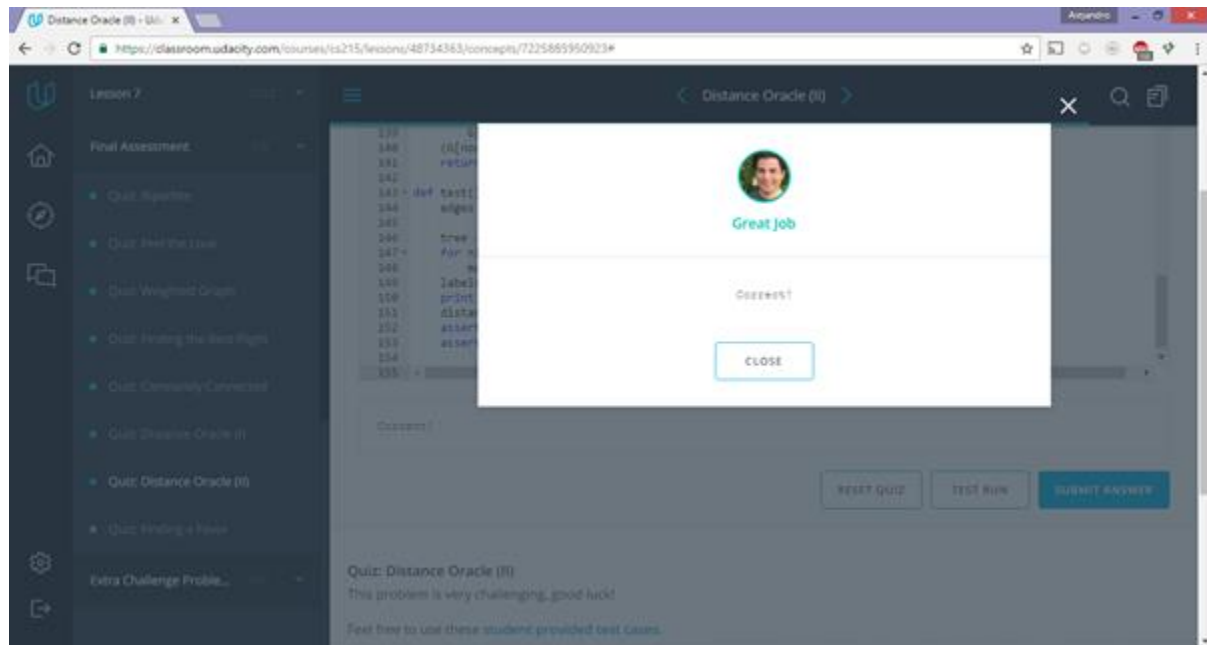
```

6. Distance Oracle (I)



```
def create_labels(binarytreeG, root):  
    labels = {root: {root: 0}}  
    frontier = [root]  
    while frontier:  
        cparent = frontier.pop(0)  
        for child in binarytreeG[cparent]:  
            if child not in labels:  
                labels[child] = {child: 0}  
                weight = binarytreeG[cparent][child]  
                labels[child][cparent] = weight  
                # make use of the labels already computed  
                for ancestor in labels[cparent]:  
                    labels[child][ancestor] = weight +  
labels[cparent][ancestor]  
                frontier += [child]  
    return labels
```

7. Distance Oracle (II)



```
def apply_labels(treeG, labels, found_roots, root):  
    if root not in labels: labels[root] = {}  
    labels[root][root] = 0  
    visited = set()  
    open_list = [root]  
    while open_list:  
        c_node = open_list.pop()  
        for child in treeG[c_node]:  
            if child in visited or child in found_roots: continue  
            if child not in labels: labels[child] = {}  
            labels[child][root] = labels[c_node][root] +  
treeG[child][c_node]  
            visited.add(child)  
            open_list.append(child)
```

```

def update_labels(treeG, labels, found_roots, root):

    best_root = find_best_root(treeG, found_roots, root)

    found_roots.add(best_root)

    apply_labels(treeG, labels, found_roots, best_root)

    for child in treeG[best_root]:

        if child in found_roots: continue

        update_labels(treeG, labels, found_roots, child)


def create_labels(treeG):

    found_roots = set()

    labels = {}

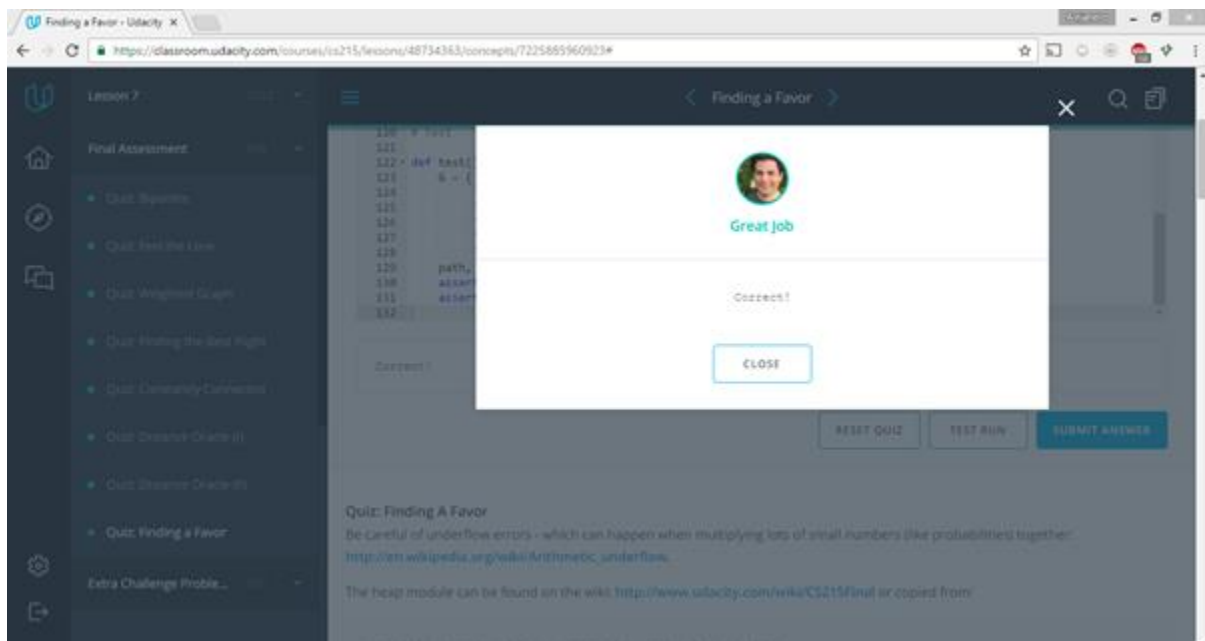
    # your code here

    update_labels(treeG, labels, found_roots, iter(treeG).next())

    return labels

```

8. Finding a Favor



```

def maximize_probability_of_favor(G, v1, v2):
    # your code here

    from math import log, exp

    logG = {}

    n = len(G.keys())

    m = 0

    for node in G.keys():
        logG[node] = {}

        m += len(G[node].keys())

        for neighbor in G[node].keys():
            logG[node][neighbor] = -log(G[node][neighbor])

    if n**2 < (n+m)*log(n):
        final_dist = dijkstra_list(logG, v1)
    else:
        final_dist = dijkstra_heap(logG, v1)

    if v2 not in final_dist: return None, 0

    node = v2

    path = [v2]

    while node != v1:
        node = final_dist[node][1]

        path.append(node)

    path = list(reversed(path))

    prob = exp(-final_dist[v2][0])

    return path, prob

```

4. Considere el problema de cubrir una tira rectangular de longitud n con 2 tipos de fichas de dominó con longitud 2 y 3 respectivamente. Cada ficha tiene un costo C_2 y C_3 respectivamente. El objetivo es

cubrir totalmente la tira con un conjunto de chas que tenga costo mínimo. La longitud de la secuencia de chas puede ser mayor o igual a n, pero en ningún caso puede ser menor.

a) Subestructura óptima

Para la resolución de un problema de longitud n, primero se obtiene la solución para una tira de longitud menor a n, calculando estas soluciones puede dar solución al problema de longitud n.

b) Ecuación recursiva.:

$$P_n = \left\{ \begin{array}{ll} \min(P_2, P_3) & \text{si } n \leq 2; \\ \min(2P_2, P_3) & \text{si } n = 3 \\ \min(P_i + P_{n-i}) & 1 \leq i < n-1 \end{array} \right. \text{ si } n > 3$$

c) Programa python:

```
def cubrir(C2, C3, n, r):
    r[0] = 0
    q = float('inf')
    if n == 1 or n == 2:
        q = min(C2, C3)
    elif n == 3:
        q = min(2 * C2, C3)
    if i in r and (n - i) in r:
        q = min(q, r[i] + r[n - i])
    else:
        q = min(q, cubrir(C2, C3, i, r) + cubrir(C2, C3, n - i, r))
    r[n] = q
    return q
```

d) Tabla para C2 = 5, C3 = 7, n = 10.

N	1	2	3	4	5	6	7	8	9	10
cubrir(5,7,n)	5	5	7	10	12	14	17	19	21	24

5. Problema de cubrimiento de un tablero 3 x n con chas de domino:

❖ Ecuaciones:

$$A_n = D_{(N-1)} + C_{(N-1)}$$

$$c_n = 2 * A_{(N-2)}$$

$$D_n = D_{(N-2)} + 2 * C_{(N-1)}$$

- B_n y E_n son siempre cero ya que no es posible que resulte la forma del tablero que representan.
- Implementación:

```
def A(N):  
    if N == 0:  
        return 0  
  
    if N <= 1:  
        return 1  
  
    return D(N - 2) + C(N - 1)
```

```
def C(N):  
    if N == 0:  
        return 0  
  
    if N <= 2:  
        return 1  
  
    return A(N - 1)
```

```
def D(N):  
    if N == 0:  
        return 0  
  
    if N <= 2:  
        return 3  
  
    return D(N - 2) + 2*A(N-1)
```

- Resultados:

10	50	100
203	238039524083	31208688988045323113527764971