```
Let min cost = 0
Let root.cost = 0
Let L be an empty set // stores left nodes for shortest path
Let R be an empty set // stores right nodes for shortest path
Let P be an empty set // represents shortest path
Let P[0] = root
Let L[0] = root
Let R[0] = root
MinimumCost(G=(V, E), P, n, p, q, root, min_cost) {
   if all edges have been visited, then
        return p * L.length()-1 + q * R.length()-1
   else if root is null {
       return
   else {
        if the left edge has not been visited, then
           mark edge as visited
           add root.left to L
            add root.left to P
            return MinimumCost(G=(V, E), P, n, p, q, root.left, min cost +
(q
        if the right edge has not been visited, then
           mark edge as visited
           Let root.right.cost = min cost + q
            add root.right to R
            add root.right to P
            return MinimumCost(G=(V, E), P, n, p, q, root.right, min cost
        if both leaves are null, then
            if min cost > root.cost, then
                for vertex in P {
                    remove vertex in L not found in P
                    remove vertex in R not found in P
                    if vertex not in R, then
                        add vertex in P not found in L to L
                    if vertex not in L, then
                        add vertex in P not found in R to R
```

```
2) LocalMinimum(f[a...b]) {
   Let m = Math.floor(a+b/2);
   if (f[m-1] > f[m] \text{ and } f[m+1] > f[m])  {
        return f[m];
   else {
        if (f[m-1] > f[m] \text{ and } f[m+1] < f[m])
            return LocalMinimum(f[m+1...b]);
       else if (f[m-1] < f[m]  and f[m+1] > f[m])
            return LocalMinimum(f[a...m]);
3) MaxBandwidth(G=(V, E), s, d) {
   Let S be an empty array // Stores shortest path
   Let V be an empty array // Stores current path
   Let max bandwidth = 0 // Represents highest bandwidth
   Let s.bandwidth = 0
   Let S[0] = s
   Relax(s)
   while d is not in S {
       Let max = 0
       Let vertex = null
       for vertex v in V {
            if max < v.bandwidth, then
                Let max = v.bandwidth
                Let vertex = v
       remove vertex from V
       add vertex to S
       Relax(vertex)
   return S
```

```
Relax(curr) {
    for v adjacent to curr that is not in S {
        Let max bandwidth = curr.bandwidth + bw(curr, v)
        if v is not in V, then
            Let v.bandwidth = max bandwidth
            Let curr = max bandwidth;
            add v to V
            Let v.bandwidth = max bandwidth
Proof:
Theorem: Dijkstra's Algorithm will find the path with the maximum
bandwidth from s to d.
Proof:
Consider an arbitrary weighted connected undirected graph G=(V, E), with
no negative bandwidths and s \in Vand d \in V1.
Let S be a list of vertices and bandwidths assigned by Dijkstra's
algorithm in the order the vertices
are relaxed by Dijkstra's algorithm. Let O be a list of vertices and their
maximum bandwidths in the same order as S.peek
Let i be the index in S and O where a disagreement occurs such that s_i =/=
o_i . Since O is optimal, then the only possible outcome is that o(i) >=
s(i) or else there is a contradiction in which the algorithm is buggy
because O always returns the path with the maximum bandwidth. Let C be the
vertices prior to v upon which S and O agree.
Consider any path v other than the one found by S. Any such path must
start in C, leave C via another vertex u and then return
to v with \overline{0} or more vertices in between.
Since u is connected to a vertex in \mathcal{C}_{m{\prime}} it must be a vertex for which we
have an estimate. Since v is the vertex with
```

maximum estimate u must have a lesser (or equal) estimate to v. Since the cost to reach u is already less than the cost to reach v, we can assume that u is not optimal because there is only one path with the maximum bandwidth. As such, if there are an arbitrary number of vertices in between u and d as there are between v and d, then only v will provide the optimal solution.

- 4) Refer to Myers.java file for results
- 5) Theorem: Select elements in A in descending order such that the maximum of a_i * $\log{(b_i)}$ is always selected first given that B is in descending order as well.

Proof: Consider an arbitrary instance of the payoff problem.

Let B[1...n] represent a set where its elements are sorted in descending order.

Let S[1...n] represent the selected set A in descending order that is used in our algorithm to produce the maximum payout.

Let O[1...n] be some optimal set of A in descending order that produces the maximum payout.

Let i be the first instance where S and O disagree on.

If $s_i = o_i$, there is no disagreement. The algorithm holds.

If s_i < o_i , then there is a contradiction because our algorithm always selects the element in A

with the highest value first provided that set B is in descending order.

If $s_i > o_i$, then consider all of the elements in O from i+1 to an arbitrary j. Because O selects elements with

the highest value, then all elements in O[i+1....j] < O[i] must be true. As such, the highest product of a_i * log(b_i) will always be chosen which makes this a valid solution.

As such, we can replace o(i) in O with s(i) to create a new solution O'
O' is valid because it will still select the elements with the highest
element in A first.

O' is optimal since the highest payout is still produced according to the algorithm.

Now 0 and S agree at point i. Repeat this argument for every i in disagreement to show that S = 0 for some optimal set 0.