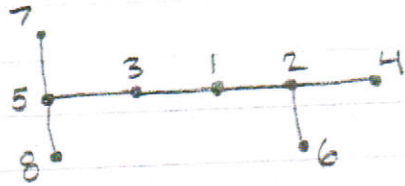
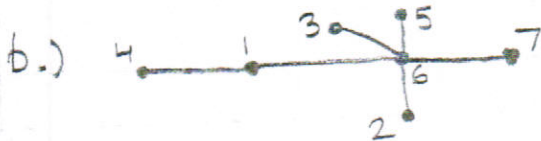


1.) i.) a.)



$$\begin{aligned} i=1, v=4, s_1=2 \\ i=2, v=6, s_2=2 \\ i=3, v=2, s_3=1 \\ i=4, v=1, s_4=3 \\ i=5, v=3, s_5=5 \\ i=6, v=7, s_6=5 \end{aligned}$$

$$\Rightarrow \boxed{\langle 2, 2, 1, 3, 5, 5 \rangle}$$



$$\begin{aligned} i=1, v=2, s_1=6 \\ i=2, v=3, s_2=6 \\ i=3, v=4, s_3=1 \\ i=4, v=1, s_4=6 \\ i=5, v=5, s_5=6 \end{aligned}$$

$$\Rightarrow \boxed{\langle 6, 6, 1, 6, 6 \rangle}$$

ii.) a.)

$$i=1, L=1, 2, 3, \textcircled{4}, 5, 6, 7, 8$$

$$P=\textcircled{2}, 2, 1, 3, 5, 5$$

$$i=2, L=1, 2, 3, 5, \textcircled{6}, 7, 8$$

$$P=\textcircled{2}, 1, 3, 5, 5$$

$$i=3, L=1, \textcircled{2}, 3, 5, 7, 8$$

$$P=\textcircled{1}, 3, 5, 5$$

$$i=4, L=\textcircled{1}, 3, 5, 7, 8$$

$$P=\textcircled{3}, 5, 5$$

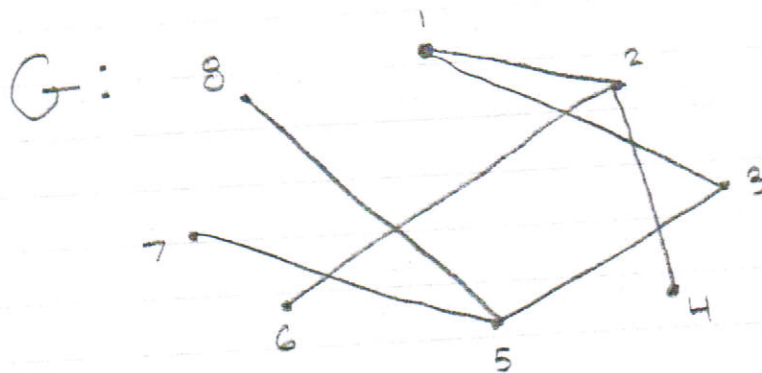
$$i=5, L=\textcircled{3}, 5, 7, 8$$

$$P=\textcircled{5}, 5$$

$$i=6, L=5, \textcircled{7}, 8$$

$$P=\textcircled{5}$$

finally we're left with $L=5, 8$, so join 5 and 8



b.)

$i=1, L=1, \textcircled{2}, 3, 4, 5, 6, 7$

$P=\textcircled{6}, 6, 1, 6, 6$

$i=2, L=1, \textcircled{3}, 4, 5, 6, 7$

$P=\textcircled{6}, 1, 6, 6$

$i=3, L=1, \textcircled{4}, 5, 6, 7$

$P=\textcircled{1}, 6, 6$

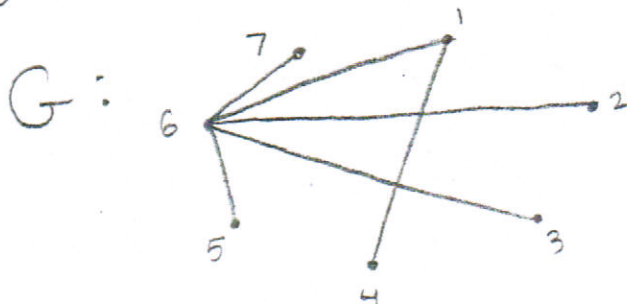
$i=4, L=\textcircled{1}, 5, 6, 7$

$P=\textcircled{6}, 6$

$i=5, L=\textcircled{5}, 6, 7$

$P=\textcircled{6}$

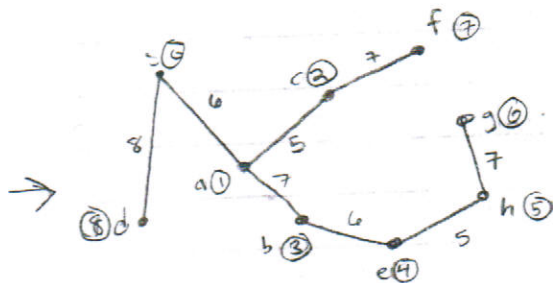
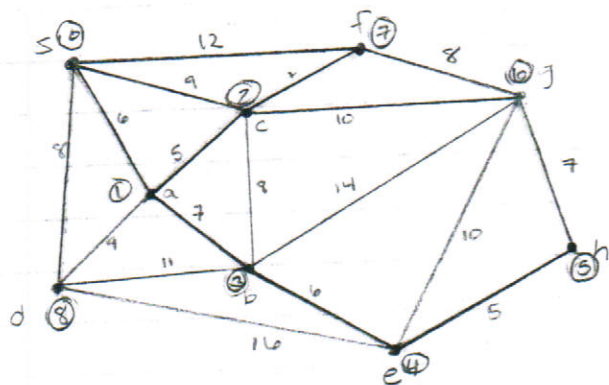
finally, we're left with $L=6, 7$, so join 6 and 7



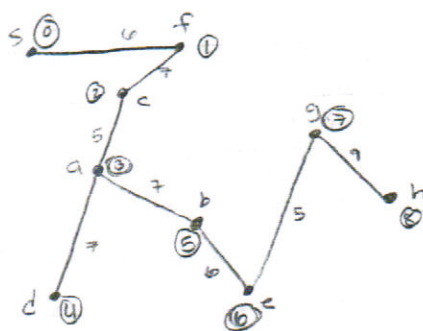
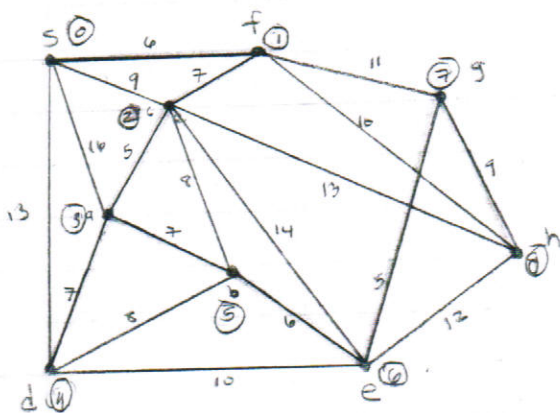
S/S

Result: The decoding process (demonstrated in ii.) is the inverse of the encoding process (demonstrated in i.).

2)

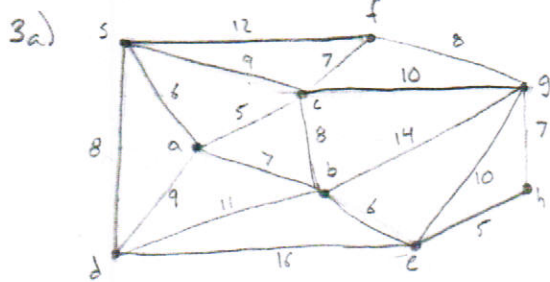


$$\sum \text{weights} = 12 + 5 + 10 + 8 + 5 + 8 + 7 = 51$$



$$\sum \text{weights} = 6 + 5 + 10 + 8 + 5 + 8 + 7 = 52$$

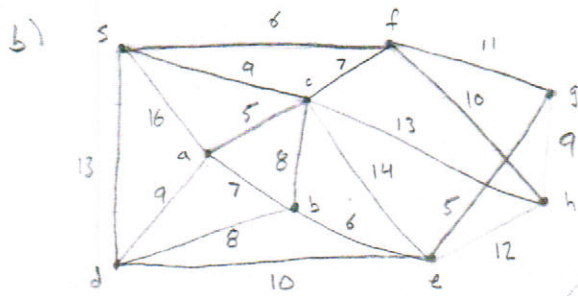
5/5



vertex: (discovery #, distance from s)

s: (0, 0)	b: (5, 13)
a: (1, 6)	e: (6, 19)
d: (2, 8)	g: (7, 19)
c: (3, 9)	h: (8, 24)
f: (4, 12)	

5/5

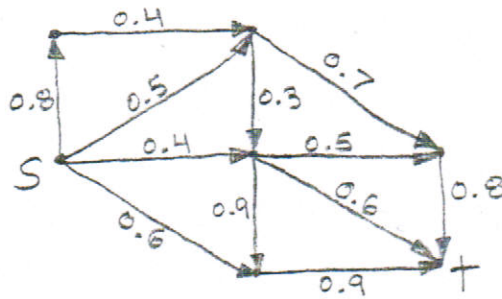


vertex (discovery #, distance from s)

s: (0, 0)	h: (5, 16)
f: (1, 6)	g: (6, 17)
c: (2, 9)	b: (7, 17)
d: (3, 13)	e: (8, 22)
a: (4, 14)	

4.)

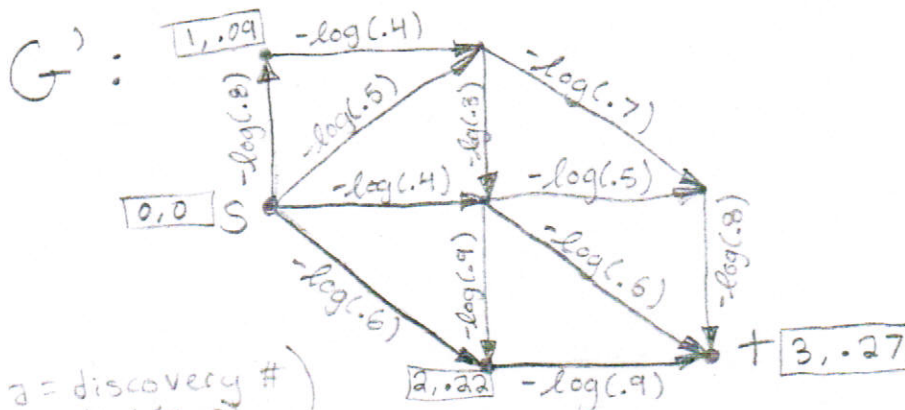
G:



edges = probability that the link does not fail (P_i)

We want to maximize the probability that the path from s to t does not fail.

So for a path $P_1, P_2, \dots, P_k$ from s to t
 $\log(P_1 P_2 \dots P_k) = \log(P_1) + \log(P_2) + \dots + \log(P_k)$
 we want to maximize $\log(P_1) + \dots + \log(P_k)$,
 but if we negate this, we'll want to minimize $-\log(P_1) - \log(P_2) - \dots - \log(P_k)$.
 Therefore, if we transform each edge P_i in G using $-\log(P_i)$, we can apply Dijkstra's algorithm. (note that for $P_i < 1$, $-\log(P_i) > 0$ so we have no negative numbers)



we can stop here!

($[a, b]$: a = discovery #
 b = dist(s, p))

$\Rightarrow$ The most reliable path from s to t is $P = s \rightarrow \text{bottom node} \rightarrow t$