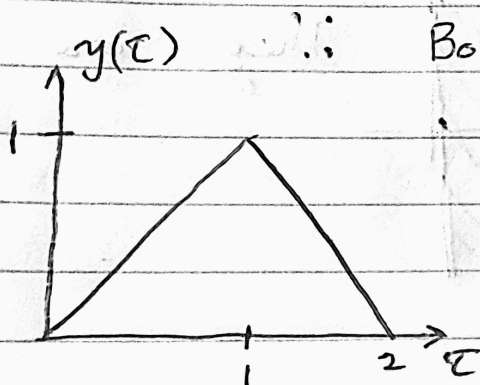
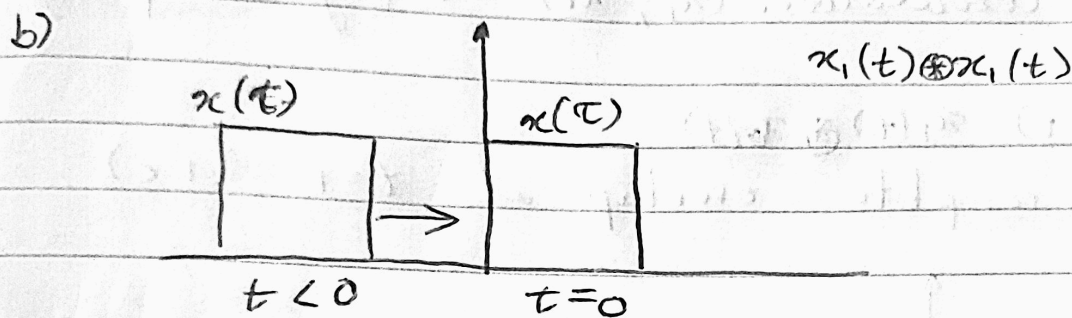
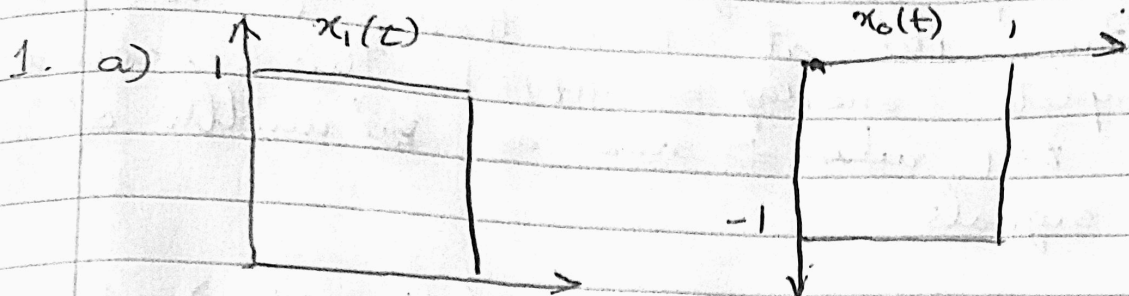


# Pre Lab



Boundaries:

$$t < 0, y(t) = 0$$

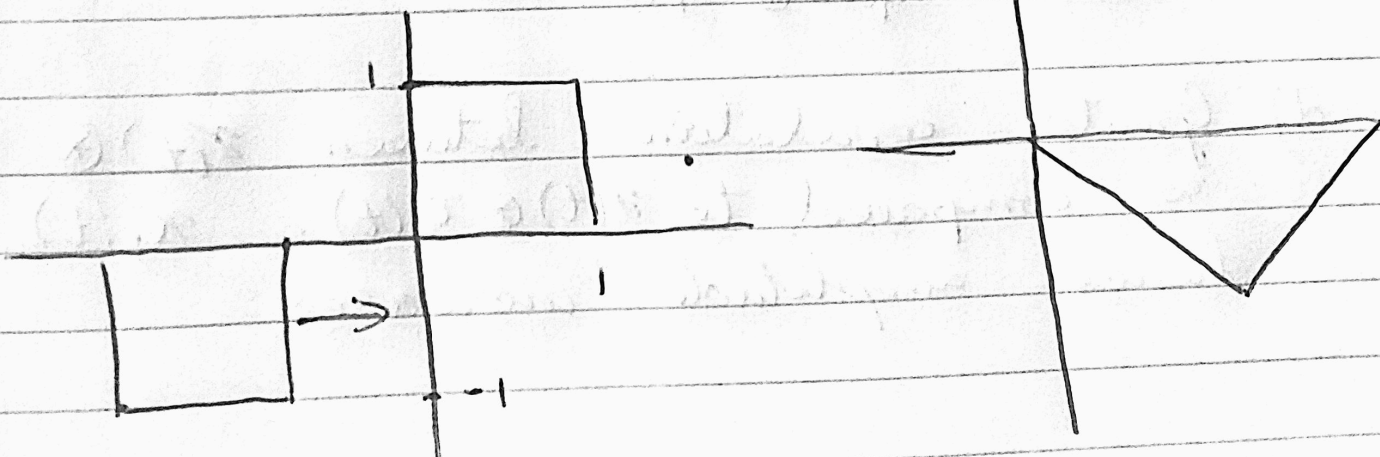
$$0 < t \leq 1, y(t) = t$$

$$1 < t \leq 2, y(t) = -t + 2$$

$$t > 2, y(t) = 0$$

Hence A is correct

c)  $x_1(t) \otimes x_0(t)$



Hence B is correct

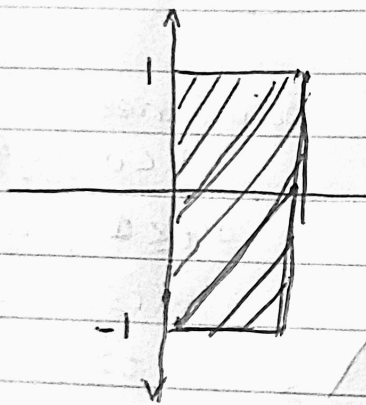
2. a) width of both signals  $w_i = 1$

b) (i) From 1b), at  $t=1$  there is a complete overlap  $\Rightarrow y(t)=1$ . This occurs at  $t=1$  ~~where~~  $\Rightarrow$  same as ~~the~~ width of ~~the~~ signals

(ii) correlation  $(x_1, x_1) = 1$  ( $y(\text{overlap}) = 1$ )

c)  $y_0(t) = x_1(t) \otimes x_0(t)$

(i) complete overlap at  $t=1$  (1.c)



They are equals.

width = 1.

(ii)  $y_0(t_{\text{overlap}}) = -1$  at  $t=1$ . This is lower than  $x_1(t) \otimes x_1(t)$

d) Greater correlation between  $x(t) \otimes x_1(t)$  as compared to  $x(t) \otimes x_0(t)$ .  $x_1(t)$  is positive hence amplitude increases

3. a) `def decode(x, d, fs, x0, x1):`  
    `high = numpy.convolve(x, x1) / fs`  
    `low = numpy.convolve(x, x0) / fs`  
    `s = numpy.zeros(int(fs * d))`

b) `for i in s:`  
    `if low[i] > high[i]:`  
        `s[i] = 1`  
    `or`  
    `return s`

4. a) `rm-rows = rm.shape[0]`

b) `message-bits = np.zeros(rm-rows, 8000)`

c) `for i in rm-rows:`

`for j in message-bits[i]:`

`message-bits[j][i] = rm[j][i]`