

① If we use $w_n = e^{2\pi q i/n}$, instead of $w_n = e^{2\pi i/n}$, we are simply raising the ~~power~~ of n^{th} roots of unity to the power of q . and so we are considering $(n/q)^{\text{th}}$ roots of unity. Although we are considering ' n ' values to compute the point-values, we are actually computing the same value ' q ' times and so, there will be only ' n/q ' distinct values.

So, if there are ' n ' distinct values with the original FFT algorithm there will be ' n/q ' distinct values with the new algorithm.

② The iterative-FFT computes the twiddle factor as many number of times as the innermost loop (lines-9-13) runs for each stage. The outer loop runs for n/m times and inner loop runs for $m/2$ times for each of outer loop run. Hence for each stage, the total number of times the twiddle factor computed are

$$n/m * m/2 = \underline{\underline{n/2}}$$

If we look at the example of iterative-fft with $n=8$, in the first stage we use $w_m = w_2$ and the inner loop runs only 1 time. So, the unique twiddle values of w are $1, w_2$. Similarly, if we look at second stage, we use $w_m = w_4$ and the inner loop runs for 2 times and so, the values of w are $1, w_4, w_4^2$. Hence, we need to compute only w_4 and w_4^2 for second stage. Similarly, for the 3rd stage $w_m = w_8$ and inner loop runs 4 times and the unique values of w are $1, w_8, w_8^2, w_8^3$. Hence computing these values before hand, will reduce the number of computations and so, at each stage s we require only 2^{s-1} twiddle factors and so, only those number of computations.

③ (a) The twiddle factor $\omega_n^{n/2}$ ~~will be~~ ^{is} computed by most number of multiplications. When the 'log n' stage is running, the ω_n will be ω_n , the n th roots of unity, and the inner loop runs for ' $n/2$ ' ^{times} each time multiplying w by ω_n and updating 'w' value is 'i'.

Hence, $\omega_n^{n/2}$ is computed by $n/2$ multiplications.

⑥

- ④ Given a function $\text{rev}_k(a)$ which runs in $\Theta(k)$ time consider, the following algorithm to compute the bit-reversal permutation on an array of input size $n=2^k$.

BIT-REVERSAL-PERMUTATION(A)

1. $n = A.\text{length}$
2. for $i = 0$ to $n-1$
3. $B[i] = 0$
4. for $i = 0$ to $n-1$
5. if $B[i] == 0$
6. $\text{reverse} = \text{rev}_k(i)$
7. $\text{temp} = A[i]$
8. $A[i] = A[\text{reverse}]$
9. $A[\text{reverse}] = \text{temp}$
10. $B[i] = 1$
11. $B[\text{reverse}] = 1$
12. end-if
13. end-for

We consider a new array B , which will ~~have~~ have '0' if a swap is not performed for that index and 1 if a swap is performed. The elements $A[i]$ needs to be swapped with $A[\text{rev}_k(i)]$ only ~~once~~ and so using this array we ensure that we swap only once.

~~Since the loop runs for i times~~

Although the loop at line 4 runs for n times, ^{will happen} swapping of ~~at most~~ $n/2$ element for $n/2$ times and so the $rev_k(a)$ function is called $n/2$ times. Hence the running time is $O(nk)$

Although a slight improvement can be found by running the loop in line 4 ~~from~~ from 1 to $n-2$, since $rev_k(0) = 0$ and $rev_k(n-1) = n-1$ but that does not have much impact ^{if n is pretty large.} on the ~~running~~ ~~average~~ ~~running~~

(b) The bit-reversed increment is pretty much similar to the bit-increment algorithm in the textbook (page-454) and the difference is that in BIT-INCREMENT, we start from lower bit, here we start from higher order bit.

BIT-REVERSED-INCREMENT(A)

1. ~~is equal~~ $n = A.length$
2. ~~while~~ ~~$i = 0$~~ and $i = n-1$
3. While $i \geq 0$ and $A[i] = 1$
4. $A[i] = 0$
5. $i = i-1$
6. end-while
7. if $i \geq 0$
8. $A[i] = 1$
9. end-if.

Grading by their approach

As we know that ~~by~~ the amortized cost of BIT-INCREMENT is constant, the amortized cost of this algorithm is also constant and so for an array of n element, we can find the reverse of bit-reversed number of an index for each in constant time and so the overall cost is $O(n)$

- © The BIT-REVERSED-INCREMENT does not use any shifting of words and so, the algorithm still we can still calculate the n -element bit-reversed permutation in $O(n)$ time.