FFT

Find  $k^{th}$  smallest element:

fxn select(A[0..n-1],k)
 if n=1, return A[0]
 pick some v=A[i]
 A< : entries < v, length n<
 A= : entries = v, length n=
 A> : entries > v, length n>
 if k < n<, select(A<,k)
 if n< <= k < n>, return v
 else select(A>,k-n<-n=)</pre>

## PARALLEL COMPUTING

Essentially, you have O(work done \* num of sequential levels + work done \* num of parallelized levels). If you have  $p \geq n$ , then you do  $\frac{f(n)}{p}$  work at each part at  $\log p$  levels. If p < n, then you have  $\frac{f(n)}{p}$  work at  $\log p$  parallelized levels +  $\frac{f(n)}{p}$  work done at  $\log \frac{n}{p}$  sequential levels.

## RECURRENCES

Given f(x), g(x): f = O(g) if  $\frac{f}{g} \leq \text{constant}$  (has upper bound). This means that  $g = \Omega(f)$ .  $f = \Theta(g)$  means that f = O(g) and g = O(f) (f and g differ by constant factor).  $n^a$  dominates  $n^b$  if a > b. Any exponential dominates any polynomial  $3^n > n^5$ . Any polynomial dominates any  $\log (n > (\log n)^3, n^2 > n \log n)$ .  $f_1 = O(g_1)$  and  $f_2 = O(g_2) \to f_1 + f_2 = O(g_1 + g_2)$  (same with multiplication).  $\log(n) = O(n)$ . if  $\frac{f(n)}{g(n)} \to \text{finite nonzero constant as } n \to \infty$ , then  $f = \Theta(g)$ . If  $\frac{f(n)}{g(n)} \to 0$  as  $n \to \infty$ , then f = O(g) but not  $\Omega(g)$ . If you have confusing exponents, try taking the log of both sides.

$$T(n) = \begin{cases} \Theta(f(n)) & \text{if } f(n) = \Omega(n^d) \quad d > \log_b a \\ \Theta(f(n)\log n) & \text{if } f(n) = \Theta(n^d) \quad d = \log_b a \\ \Theta(n^{\log_b a}) & \text{if } f(n) = O(n^d) \quad d < \log_b a \end{cases}$$

Compare f(n) to  $n^{\log_b a}$ . This lets you know which one of these cases you want to choose d to fit.

## SUMMATIONS

$$\sum_{i=0}^{n} = \frac{n(n+1)}{2} \sum_{i=0}^{n} i^2 = \frac{n(n+1)(2n+1)}{6} \sum_{k=0}^{n-1} ar^k = a\frac{1-r^n}{1-r} \sum_{i=1}^{\log n} n = n\log n \sum_{i=1}^{n} \frac{1}{2^i} = 1 - \frac{1}{2^n}$$

When doing FFT, you evaluate your polynomial at the  $n^{th}$  roots of unity up to the closest power of 2:  $\{1, i, -1, -i\}$  for  $F_4$ . Do do the *inverse* FFT, evaluate it at the complex conjugate of that same vector:  $\{1, -i, -1, i\}$  for  $F_4$ . We do y = Fx to change polynomial from coefficients x to values y.

fxn FFT(A,w):

in: coefficient representation of polynomial

A(x) of degree <= n-1, w=nth root of unity
out: value representation A(w^0)..A(w^n-1)
if w=1, return A(1)
express A(x) in form A\_e(x^2) + xA\_0(x^2)

FFT(A\_e,w^2) to evaluate A\_e at even powers of w

FFT(A\_o,w^2) to evaluate A\_o at even powers of w

for j=0..n-1:

compute A(w^i) = A\_e(w^2i)+w^i\*A\_o(w^2i)

compute  $A(w^j) = A_e(w^2j)+w^j*A_o(w^2j)$ return  $A(w^0)...A(w^n-1)$ 

FFT example on  $P(x) = 2x^0 + 3x^1 + 4x^2 + 5x^3$ . Rewrite as even and odd terms:  $P(x) = (2x^0 + 4x^2) + x(3x^0 + 5x^2)$ . To make a recursive call here, we need to have P(x) with degree < 2, because we're moving from  $F_4$  to  $F_2$ . We divide the exponents of  $P_{\text{even,odd}}$  by 2, then plug in  $x^2$  to preserve the polynomial.  $P(x) = P_{\text{even}}(x^2) + xP_{\text{odd}}(x^2)$ . Evaluate at the  $2^{th}$  roots of unity 1, -1, then we can evaluate P(1, i, -1, -i) by using the fact that each of those squared is either 1 or -1.

## GRAPHS

To find source, run DFS and the vertex w/ largest post[v] is in a source SCC. To find sink, do the same process on same graph w/ edges reversed. The source in  $G_R$  will be a sink in G.