



Given that available algorithm for Steiner tree is of complexity  $2^n$  ( $|F| = n$ )  
 So we need  $\log_2 n$  points per division  
 to get polynomial complexity ( $n^{o(1)}$ )

Expected no. of points per square ( $\frac{1}{m} \times \frac{1}{m}$ )

$$\Rightarrow \frac{n}{m^2} = \log n$$

$$m = \left( \frac{n}{\log n} \right)^{1/2}$$

Algorithm:

Step 1: Divide  $[0,1]^2$  in  $m^2$  small squares.

Each square will give  $G_i, F_i$

(partitions of  $G, F$  present in the square)

Apply our  $2^n$  algorithm to  $G_i, F_i$ .

Step 2: (Combining problems)

we have got ST in each small square,  
 now we start from left bottom square

and join any leaf in it to any leaf in the square right to it. we will continue like this in snake fashion (shown in diagram) until top right/left square. Since we are joining leaves tree property remains intact. This combining process may not be optimal but the cost incurred will be well within the bounds.

$$\begin{aligned} \text{Cost of combining} &\leq m^2 o(\text{diameter of square}) \\ &= o(m^2 \times \frac{1}{m}) \\ &= o(m) \quad (\text{Same as TSP}) \end{aligned}$$

$$\begin{aligned} \text{Total cost} &= \sum_{i=1}^{m^2} ST(F_i, G_i) + \text{cost of combining} \\ &= \sum_{i=1}^{m^2} ST(F_i, G_i) + o(m) \quad \text{--- (1)} \end{aligned}$$

b) Expected running time:

$$\Rightarrow m^2 \underbrace{o(2^{\frac{n}{m^2}})}_{\substack{\text{Expected} \\ \text{time for each square}}} + o(m^2)$$

$$\Rightarrow \frac{n}{\log n} 2^{\frac{n}{n} \log n} + \frac{n}{\log n}$$

$$\Rightarrow o\left(\frac{n^2 + n}{\log n}\right) \approx o\left(\frac{n^2}{\log n}\right)$$

Cost computation:

\* Growth bound for  $ST(F, G)$  [ $G \subseteq F$ ]

→ Since  $ST(F, G)$  is euclidean and subadditive it satisfies growth bound

$$ST(F, G) \leq o\left(n^{\frac{d-1}{d}}\right) \leq o(\sqrt{n}) \text{ --- (2)}$$

in expectation

\* Bound on  $ST_B(F, G)$

By definition

$$ST_B(F, G) \leq ST(F, G)$$

from (2)

$$ST_B(F, G) \leq o(\sqrt{n}) \text{ --- (3)}$$

Take (1)

$$\Rightarrow PT(x) \leq \sum_{i=1}^{m^2} ST(F_i, G_i) + o(m)$$

→ By point wise closeness which is implication of subadditivity, smoothness and boundary functional.

$$ST(F_i, G_i) \leq ST_B(F_i, G_i) + \frac{1}{m} C |F_i|^{\frac{d}{d-1}}$$

→ factor  $1/m$  is due to reduce square dimensions

$$\Rightarrow PT(x) \leq \sum_{i=1}^{m^2} (ST_B(F_i, G_i) + \frac{1}{m} C |F_i|^{\frac{d}{d-1}}) + o(m)$$

$$\Rightarrow PT(x) \leq \sum_{i=1}^{m^2} ST_B(F_i, G_i) + \sum_{i=1}^{m^2} \frac{1}{m} C |F_i|^{\frac{d}{d-1}} + o(m)$$

→ w.k.T boundary functionals are super additive

$$\Rightarrow PT(n) \leq ST_B(F, G) + \sum_1^{m^2} \frac{c}{m} + o(m)$$

→ from (3)

$$\Rightarrow PT(n) \leq o(\sqrt{n}) + \sum_1^{m^2} \frac{c}{m} + o(m)$$

$$\Rightarrow PT(n) \leq o(\sqrt{n}) + m^2 \frac{c}{m} + o(m)$$

$$\Rightarrow PT(n) \leq o(\sqrt{n}) + o(m)$$

→ substituting  $m = \sqrt{n}(\log n)$

$$\Rightarrow PT(n) \leq o(\sqrt{n} + \sqrt{n}/(\log n)) \text{ in expectation}$$

$$\boxed{E(PT(n)) \leq o(\sqrt{n} + \sqrt{n}/(\log n))} \rightarrow (4)$$

\* Given that  $ST(F, G)$  satisfies limit theorem

$$\lim_{n \rightarrow \infty} \left( \frac{ST(F, G)}{n^{1/2}} \right) = \alpha \quad \text{almost surely } pr = 1 - o(1)$$

So we can say for sufficiently large  $n$

$$E(ST(F, G)) \leq o(n^{1/2}) \rightarrow (5)$$

from (4) & (5)

$$\Rightarrow \frac{E(PT(x))}{E(ST(x))} \leq O\left(\frac{\sqrt{n}}{\sqrt{n}} + \frac{\sqrt{n/\log n}}{\sqrt{n}}\right)$$

$$\Rightarrow \boxed{\frac{E(PT(x))}{E(ST(x))} \leq O\left(1 + \frac{1}{\sqrt{\log n}}\right)}$$

c) Time complexity =  $O\left(\underbrace{\sum_{i=1}^{m^2} 2^{n_i}}_{\text{each square}} + \underbrace{m^2}_{\text{combing}}\right)$

$$\Rightarrow E(TC) = E\left[\sum_{i=1}^{m^2} 2^{n_i} + m^2\right]$$

$$\rightarrow E[X+Y] = E[X] + E[Y]$$

$$\Rightarrow \sum_{i=1}^{m^2} E(2^{n_i}) + E(m^2) = E(TC)$$

$\rightarrow$   $m$  value doesn't change with perturbation

$$\Rightarrow m^2 + \sum_{i=1}^{m^2} E[2^{n_i}] = E(TC)$$

$\rightarrow$  let  $\gamma = [0,1]^2$  and  $s$  is small square after division

$$f(\gamma) \leq \phi \rightarrow f(s) \leq \phi/m^2$$

$$\text{w.k.t } f(s) = \Pr[x \in s] = \phi/m^2 \rightarrow (6)$$

$$\rightarrow \text{from (6)} \quad E[n_i] = \sum_{i=1}^n \Pr[n \in s_i]$$

$$E[n_i] = n\phi/m^2 \rightarrow (7)$$

$\rightarrow 2^n$  is convenient hence we can say  $E[2^n] \geq 2^{E(n)}$  but this is not useful as  $\phi$  is bounded ( $\phi < 1$ ) we can safely assume that

$$E[2^{n_i}] = O(2^{E(n_i)})$$

(for some large constant)

$$\begin{aligned}
 \Rightarrow E(\tau_c) &\leq m^2 + O\left(\sum_{i=1}^m 2^{E(n_i)}\right) \\
 &\leq m^2 + \sum_{i=1}^m 2^{n\phi/m^2} \\
 &\leq m^2 + m^2 2^{n\phi/m^2}
 \end{aligned}$$

$\rightarrow$  substituting  $m = \sqrt{n/\log n}$

$$\begin{aligned}
 E(\tau_c) &= O\left(\frac{n}{\log n} + \frac{n}{\log n} 2^{\phi \sqrt{n/\log n}}\right) \\
 &= O\left(\frac{n}{\log n} + \frac{n \cdot n^\phi}{\log n}\right)
 \end{aligned}$$

$$E(\tau_c) \simeq \frac{n^{(\phi+1)}}{\log n}$$