

~~Lecture 4.3~~ Lecture 5.1

Recursive relations

$$(1) \quad R_{a,b}(d) = \frac{da}{db} \cdot \frac{1 - \text{Bad}_{a,b}(d - e_a)}{1 - \text{Bad}_{b,a}(d - e_b)} \quad \text{for odd } d$$

$$(15) \quad \text{Bad}_{a,b}(d) = \frac{1}{da} \left(P_{ab}(d) + \sum_{v \neq a,b} Y_{avb}(d) \right) \quad \text{even } d$$

$$(2) \quad P_{av}(d) = dv \left(\sum_{b \neq v} R_{(b,a)}(d - e_v) \frac{1 - P_{bv}(d - e_a - e_v)}{1 - P_{av}(d - e_a - e_v)} \right)^{-1}$$

$$(3) \quad Y_{avb}(d) = \frac{P_{av}(d) (P_{bv}(d - e_a - e_v) - Y_{avb}(d - e_a - e_v))}{1 - P_{av}(d - e_a - e_v)}$$

Assuming things are well-defined (e.g. realisable in $P_{av}(d)$).

Start with $P_{av}(d) = O(\frac{d}{n}) \ll 1$.

→ Iterate applying (3), (15), (1); (2)

& get better & better approximations of $R_{a,b}(d)$.

What's the improvement after 1 round of iterations?

after 1st round $R_{a,b}(d) = \frac{da}{db} (1 + O(\frac{d}{n}))$

2nd $R_{a,b}(d) = \frac{da}{db} (1 + \text{sth.} + O(\frac{d^2}{n^2}))$

3rd round $\frac{da}{db} (1 + \text{sth. more complicated} + O(\frac{d^3}{n^3}))$

?

k rounds $R_{a,b}(d) = \text{Formula} (1 + O((\frac{d}{n})^k))$

For which d is $(\frac{d}{n})^k$ precise enough? $(\frac{d}{n})^k = O\left(\frac{1}{n^{\frac{1}{\log \log n}}}\right)$

$$\Leftrightarrow d \ll n^{\frac{k-1}{2k+1}} (\log n)^{1/2}$$

$$\Leftrightarrow d \ll n^{1 - \frac{2}{2k+1}} (\log n)^{1/(2k+1)}$$

In theory: $\forall \varepsilon > 0$, $d = n^{1-\varepsilon}$, need $O(1)$ iterations to obtain desired accuracy.

Problem 1: formulae get very complicated very quickly
(try to do 3rd iteration on paper).

Problem 2: that does not get us up to $d = \Theta(\frac{n}{\log n})$
or even $d = \varepsilon n$.

Operators

$T(\cdot)$

on functions

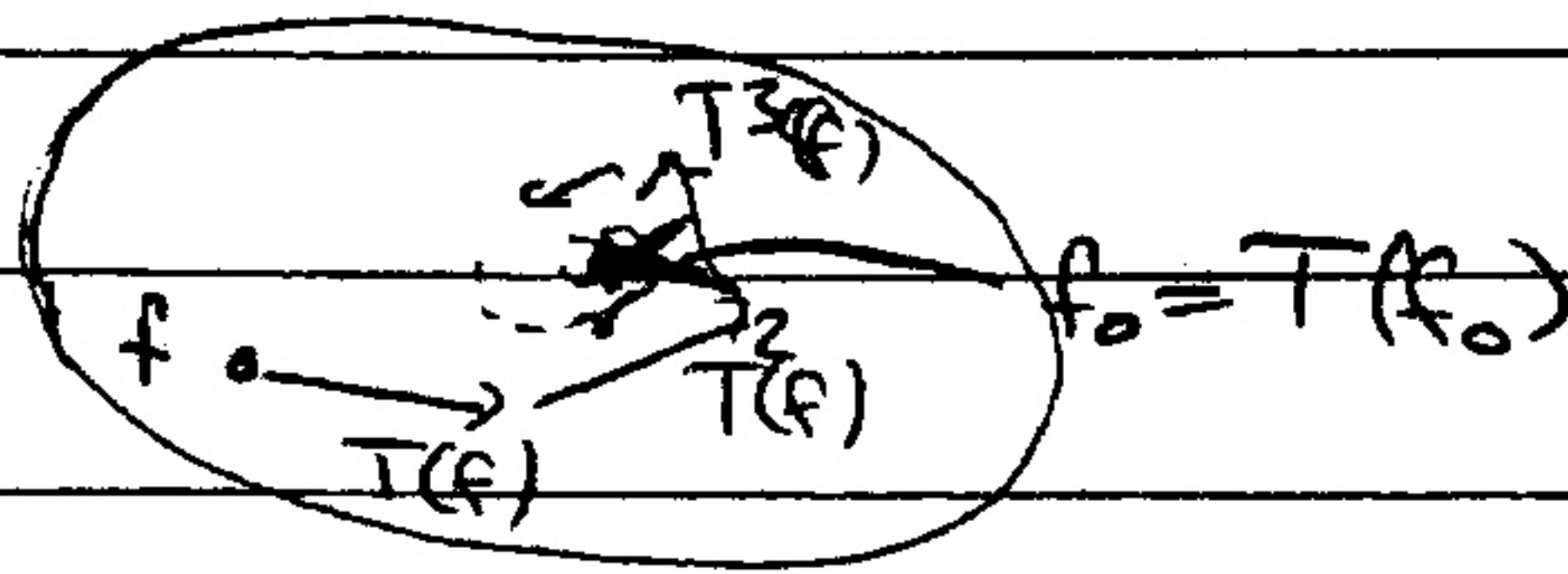
$$p = p(a, b, d)$$

$$p \in \binom{[n]}{2} \times \mathbb{Z}_{\geq 0}^n \rightarrow \mathbb{R}_{\geq 0}$$

$$p_{av}(d)$$

$$y: \binom{[n]}{2} \times \mathbb{Z}_{\geq 0}^n \rightarrow \mathbb{R}_{\geq 0} \quad y(a, v, b, d) \quad y_{avb}(d)$$

$$r: [n]^2 \times \mathbb{Z}_{\geq 0}^n \rightarrow \mathbb{R}_{\geq 0} \quad r(a, b, d) \quad r_{ab}(d)$$



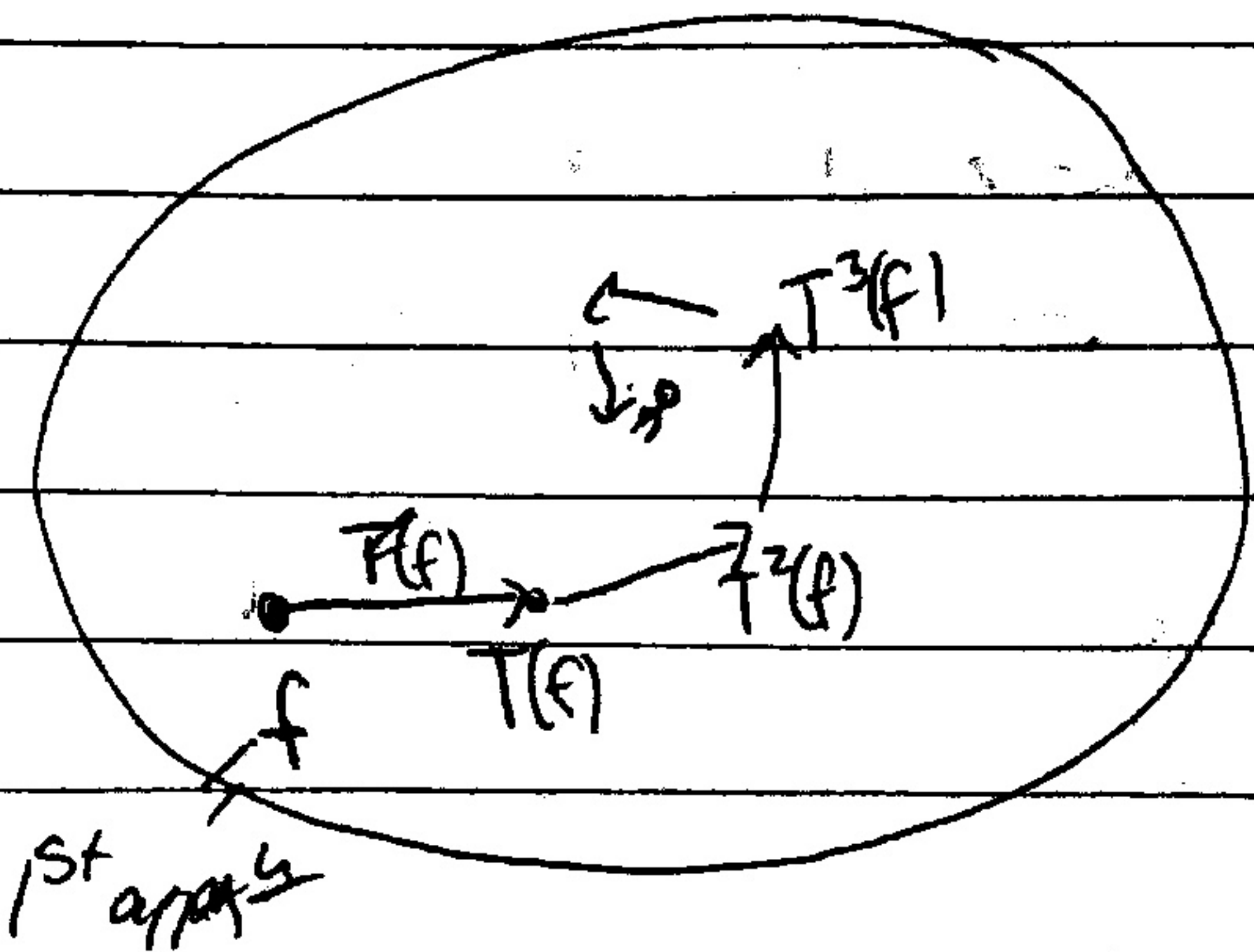
$$(1) \quad R(p, y)(a, b, d) = \frac{d_a}{d_b} \cdot \frac{1 - \text{bad}(p, y)(a, b, d - e_b)}{1 - \text{bad}(p, y)(a, b, d - e_a)},$$

$$(2) \quad \text{bad}(p, y)(a, b, d) = \frac{1}{d_a} \left(\cancel{p_{ab}(d)} + \sum_{v \neq a, b} y_{avb}(d) \right)$$

$$(2) \quad P(p, r)(a, v, d) = d_v \left(\sum_{b \neq v} r_{bva}(d - e_v) \frac{1 - p_{bv}(d - e_a - e_v)}{1 - p_{av}(d - e_a - e_v)} \right)^{-1}$$

$$(3) \quad y(p, y)(a, v, b, d) = \frac{p_{av}(d) (p_{bv}(d - e_a - e_v) - y_{avb}(d - e_a - e_v))}{1 - p_{av}(d - e_a - e_v)}$$

Observation: "real" probabilities & ratios p_{av} , y_{avb} & r_{aib} are "fixed points"



$$R(P, Y) = R$$

$$P(P, R) = P$$

$$y(P, Y) = Y.$$

Lecture 4.2 5.2

Operators are "contractive" for probability-like P & Y :

Call (P, Y) μ -probability-like on $\mathcal{D}_0 \leftarrow \text{some set} \subseteq \mathbb{Z}_{\geq 0}^n$

$$(P1) \quad 0 \leq P_{av}(d) \leq \mu \quad \forall d \in \mathcal{D}_0$$

$$(P2) \quad \sum_{a \neq b} Y_{avb}(d) \leq \mu \cdot d_a \quad \forall d \in \mathcal{D}_0, a, b \in [n] \text{ distinct.}$$

$$(P3) \quad 0 \leq Y_{avb}(d) \leq \mu \cdot P_{av}(d) \quad \forall d \in \mathcal{D}_0 \text{ \& } a, v, b \in [n] \text{ distinct.}$$

Exercise: Show that $\text{real}(P, Y)$ is μ -probability like

on $\mathcal{B}_{16 \log n}(\mathcal{D}_m)$, assuming $n \log^3 n \leq m = o(n^2)$

where $\mu = 5 \cdot \frac{m}{\binom{n}{2}}$.

Hint: Use Simple Switching lemma & Recursion for Y_{avb} .

Contraction lemma $(P, Y), (P', Y')$ are μ -probability like on $\mathcal{D}_0$

(a) If $P \equiv P'(1 \pm \xi)$ & $Y \equiv Y'(1 \pm \xi)$ on $\mathcal{B}_1(\mathcal{D}_0)$ r, r' fctⁿs., $\xi > 0$
then $R(P, Y) = R(P', Y')(1 + O(\mu \cdot \xi))$ on $\mathcal{D}_0$

(b) If $P \equiv P'(1 \pm \xi)$ & $r = r'(1 \pm \mu \xi)$ on $\mathcal{B}_1(\mathcal{D}_0)$
then $P(P, r) = P(P', r')(1 + O(\mu \xi))$ on $\mathcal{D}_0$

(c) If $P = P'(1 \pm \mu \xi)$ and $Y = Y'(1 \pm \xi)$ on $\mathcal{B}_2(\mathcal{D}_0)$
then $Y(P, Y) = Y(P', Y')(1 + O(\mu \xi))$ on $\mathcal{D}_0$

[w/o proof] e.g. $P(P, r) = dv \left(\sum r'(1 \pm \mu \xi) \cdot \frac{1 - P'(1 \pm \xi)}{1 - P'(1 \pm \xi)} \right)^{-1}$

Take $(P, Y, r) = (P, Y, R) = dv \left(\sum r' \frac{1 - P' + O(\mu \xi)}{1 - P' + O(\mu \xi)} (1 + O(\mu \xi)) \right)^{-1}$
and (P', Y', r') are guessed formulae. $= P(P', Y')(1 + O(\mu \xi))$

$$P_{av}^g(d) := \frac{d_a d_v}{d(n-1)} \left(1 - \frac{(d_a-d)(d_v-d)}{d(n-1-d)} \right)$$

$$R_{ab}^g(d) := \frac{d_a(n-d_b)}{d_b(n-d_a)} \left(1 + \frac{d_a-d_b}{d^2 n^2} \sum_i (d_i-d)^2 \right)$$

$$Y_{avb}^g(d) := P_{av}^g(d) P_{bv}^g(d-e_a-e_v) \left(1 + \frac{1}{n} \right)$$

→ also μ -probability like on $B_{16 \log n}(D_m)$

Lemma (Guessed formula close to a fixed point)

$$(a) \quad R(P^g, Y^g) = R^g (1 + O(\frac{\gamma}{n})) \quad \gamma = \frac{1}{dn} + \frac{\sqrt{\log n} d}{n^2}$$

$$(b) \quad P(P^g, R^g) = P^g (1 + O(\frac{\gamma}{n}))$$

$$(c) \quad Y(P^g, Y^g) = Y^g (1 + O(\frac{\gamma}{n}))$$

on $B_{16 \log n}(D_m^*)$: $D_m^* = \{d \in D_m : \sum_i (d-d_i)^2 \leq 2dn\}$ & assuming $n(\log n) \leq m \leq o\left(\frac{n}{(\log n)^{3/4}}\right)$.

Now, $P \equiv P^g (1 \pm \xi)$ & $Y = Y^g (1 \pm \xi)$ for $\xi = 1$ on D_m

Apply Contraction lemma:

$$R = R(P, Y) = R(P^g, Y^g) (1 + O(\mu))$$

$$= R^g (1 + O(\mu + \frac{\gamma}{n}))$$

$$P = P(P, R) = P(P^g, R(P^g, Y^g)) = P(P^g, R^g) (1 + O(\mu + \frac{\gamma}{n}))$$

$$= P^g (1 + O(\mu + \frac{\gamma}{n}))$$

$$Y = Y(P, Y) = Y(P^g, Y^g) (1 + O(\mu + \frac{\gamma}{n}))$$

$$= Y^g (1 + O(\mu + \frac{\gamma}{n}))$$

1 round of application $P = P^g (1 \pm \xi)$ & $Y = Y^g (1 \pm \xi)$

to $P = P^g (1 + O(\mu + \frac{\gamma}{n}))$ & $Y = Y^g (1 + O(\mu + \frac{\gamma}{n}))$

do this $\log n$ times.

$$\hookrightarrow P = P^g (1 + O(\mu + \frac{\gamma}{n})^{\log n} + O(\frac{1}{n})) \text{ & } Y = Y^g (1 + O(\mu + \frac{\gamma}{n})^{\log n} + O(\frac{1}{n}))$$

$$\mu = \frac{1}{dn}$$

$$\mu^{\log n} = o\left(\frac{1}{dn}\right)$$

$$\rightarrow R = R^g (1 + o\left(\frac{1}{dn}\right) + \gamma) \text{ on } B_1(D_m^*)$$

Remark: wider degree spread $D_m^* = \{d : \sum_i d_i = 2m \text{ & } |d_i - d| \leq d^{1/2 + \varepsilon}\}$ fixed $\varepsilon > 0$

& $d = \varepsilon n$

→ need more precise guesses & Maple assisted calculation.

Lecture 5.3

Methods for counting dense d -regular graphs - $d = O(\sqrt{\frac{n}{\log n}})$

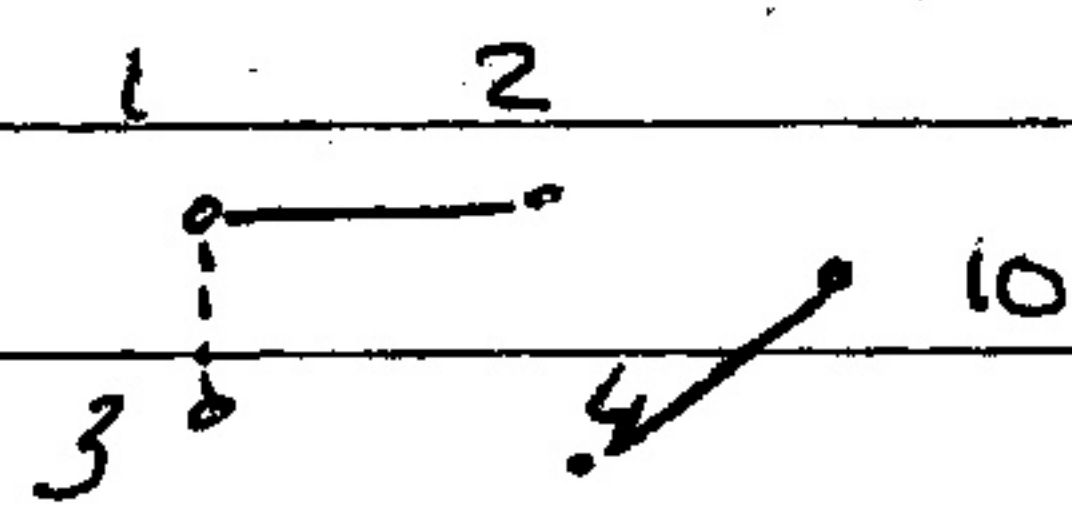
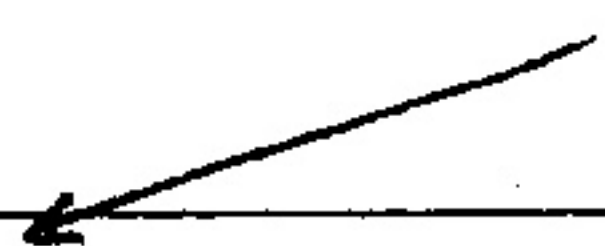
Consider $f(x_1, \dots, x_n) := \prod_{1 \leq i < j \leq n} (1 + x_i x_j)$

Expand : $2^{\binom{n}{2}}$ monomials, each corresponding to a graph on $[n]$

$(1 + x_1 x_2) (1 + x_1 x_3) \dots (1 + x_4 x_{10}) \dots$



$x_1 x_2 \cdot 1 \cdot x_4 x_{10}$



Given a monomial $x_1^{\alpha_1} x_2^{\alpha_2} \dots x_n^{\alpha_n} \rightarrow \alpha^i = d(vx^i)$ in corresponding graph

$\rightarrow$ the coefficient of $x_1^d x_2^d \dots x_n^d = \#$ labelled d -regular graphs on $[n]$.

$$= |g(n, d)|$$

Cauchy's integral formula

If $f : \mathbb{C} \rightarrow \mathbb{C}$ holomorphic (ours f is a polynomial, so ∞ th diff'ble around every pt of $\mathbb{C}$).

then $f(z)$ for U open nbhd around 0, then

$$f(0) = \frac{1}{2\pi i} \oint_{\gamma} \frac{f(z)}{z} dz$$

$$f^{(n)}(0) = \frac{n!}{2\pi i} \oint_{\gamma} \frac{f(z)}{z^{n+1}} dz \quad (1)$$

holomorphic $\Rightarrow$ analytic $\Rightarrow$ ~~power series~~

$$f(x) = f(0) + f'(0)x + \frac{f''(0)}{2!}x^2 + \dots + \frac{f^{(n)}(0)}{n!}x^n \quad (2)$$

& multivariate version

$$(1) \& (2) \Rightarrow [x^a] f(x) = \frac{1}{2\pi i} \oint_{\gamma} \frac{f(z)}{z^{a+1}} dz.$$

multivariate version

$$|g(d,n)| = [x_1^d \dots x_n^d] \prod_{k=1}^n (1+x_k x_j) = \frac{1}{(2\pi i)^n} \oint \dots \oint \frac{\prod_{j < k} (1+z_j z_k)}{z_1^{d+1} z_2^{d+1} \dots z_n^{d+1}} dz_1 dz_2 \dots dz_n$$

→ they [McKay & Wormeld] chose circles of radius $r = \sqrt{\mu/(1-\mu)}$

$$dz = i r e^{i\theta} d\theta$$

$$\dots = \frac{1}{(2\pi r^d)^n} \int_{-\pi}^{\pi} \dots \int_{-\pi}^{\pi} \frac{\prod_{j < k} (1+r^2 e^{i(\theta_j + \theta_k)})}{e^{id(\theta_1 + \dots + \theta_n)}} d\theta_1 d\theta_2 \dots d\theta_n$$

[saddle point method]

try to find apt. where integrand is largest
"saddle"

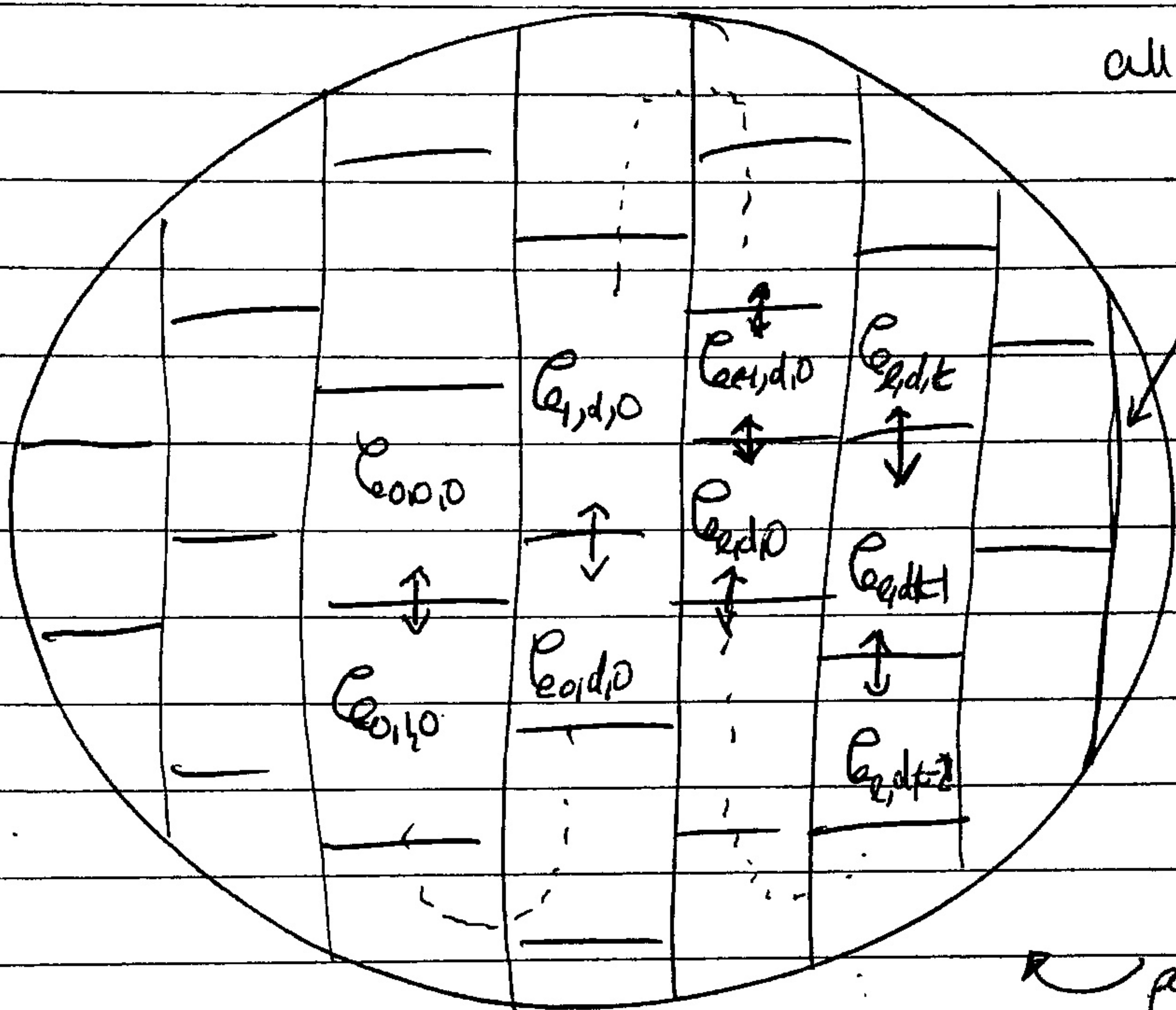
zoom in & show that, for large n , the contrⁿ from around that saddle point dominates the integral.

instead of integrating $\frac{\prod(\dots)}{e^{\dots}}$ use quadratic appⁿ around saddle point.

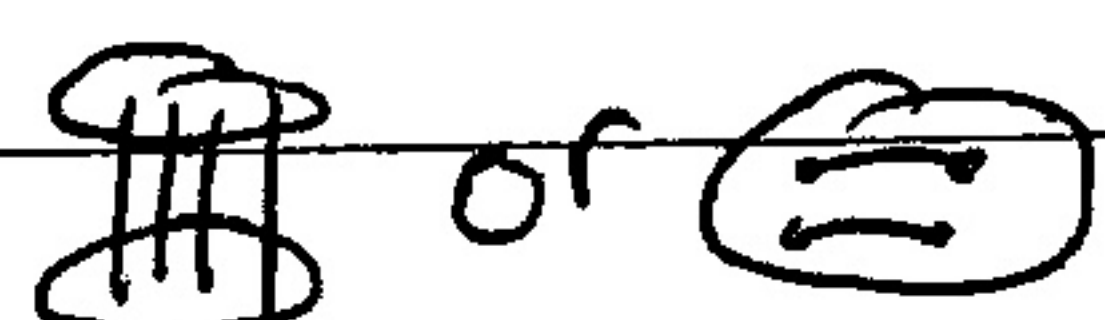
Lecture 5.4 (in case there is time)

For $\Delta^3 = o(n)$ or $d^2 = o(n)$ Recall configuration in

all pairings.



Step 1
very few have



or $> \log n$

or $> \log n$

or $> \log n$

partition the rest

into $\leq \log^3 n$ classes

$c_{e,d,t}$ — exactly e loops
d doubles

& t triples

$$\Pr(\text{Simple}) = \frac{|c_{0,0,0}|}{\sum |c_{e,d,t}|}$$

$$\frac{|c_{e,d,t}|}{|c_{0,0,0}|} = \frac{|c_{e,d,t}|}{|c_{e,d,t-1}|} \cdots \frac{|c_{e,d,1}|}{|c_{e,d,0}|} \cdot \frac{|c_{e,d,0}|}{|c_{e-1,d,0}|} \cdots \frac{|c_{e,d,0}|}{|c_{e,0,0}|} \cdot \frac{|c_{e,0,0}|}{|c_{0,0,0}|}$$

switch away
exactly 1 triple edge

switch away
exactly 1 double
edge

