

# Differential Equations Method

Consider the following simple process:

We start with  $n$  isolated vertices

$1, 2, \dots, n$ .

At a general step, we choose a (still)

isolated vertex  $v$  and add an edge

to a randomly chosen  $w$ .

Question: how long before there are no isolated vertices?

Let

$X(t)$  = # isolated vertices  
after  $t$  steps.

$$X(0) = n$$

$$\mathbb{E}(X(t+1) - X(t) \mid X(t)) = -1 - \frac{X(t)}{n-1}. \quad (*)$$

Now put  $t = \tau n$ ,  $0 \leq \tau \leq 1$

and  $n x(\tau) = X(t)$ .

(\*) on p2 suggest that

$$x'(\tau) = -1 - x(\tau)$$

given

$$x(\tau) = 2e^{-\tau} - 1.$$

In which case we would expect that  
the process ends when  $t \approx n \ln 2$ .

We now consider the following greedy algorithm for finding an independent set in a graph.

GREEDY

begin

$I \leftarrow \emptyset; A \leftarrow V;$

While  $A \neq \emptyset$  do

Choose  $v \in A;$

$I \leftarrow I \cup \{v\}; A \leftarrow A \setminus (\{v\} \cup N(v))$

[Random Choice]

end <sup>od</sup> <sub>output</sub>  $I$

Greedy produces an independent set.  
We begin by studying the likely  
size of the output, if  $G$  is a random  
 $r$ -regular graph.

We use the configuration model  
of  $r$ -regular graphs i.e.  $W = W_1 \cup W_2 \cup \dots \cup W_n$   
where  $W_i = [(i-1)r+1, ir]$

We will expose the random pairing of  $W$  as the algorithm progresses i.e. not before.

If vertex  $i$  is placed in the independent set  $I$ , then and only then, do we expose the pairs involving  $W_i$ .

Let the **degree** of a vertex  $i$  at a general step of the algorithm be the number of exposed pairs involving  $W_i$ .

Thus a general step of GREEDY involves

- (i) Choose a vertex of degree **zero**.
- (ii) Expose the pairs involving  $w_i$ .

Let  $t = |I|$  be the number of steps taken so far and let  $P_t$  refer to the current set of exposed pairs.

Let  $X(t)$  be the number of vertices of degree zero.

The number of vertices in the set chosen by GREEDY is  $t_0$ , where  $X(t_0) = 0$ .

$$E(X(t+1) - X(t) | \mathcal{P}_t) =$$

$$-1 - \frac{X(t)r}{n-2t} + O\left(\frac{1}{n}\right) \quad (*)$$

$v \in I$  (arrow pointing to  $-1$ )  
 assuming  $t \leq (\frac{1}{2} - \alpha)n$  (arrow pointing to  $O(\frac{1}{n})$ )

We expose  $r$  pairs associated with  $v$ .  
 For first pair there are still  $r(X(t)-1)$  points  
 associated with vertices of degree zero,  
 (excluding  $v$ ). There are  $r(n-2t)$  points unpaired  
 altogether. So the probability of pairing  
 with vertex of degree zero is  $\frac{r(X(t)-1)}{r(n-2t)-1} = \frac{X(t)}{n-2t} + O(\frac{1}{n})$ .

Repeat  $r$  times to get (\*).



Putting  $t = \tau n$  and  $X(t) = nx(\tau)$ , this suggests that we solve

$$x'(\tau) = -1 - \frac{rx(\tau)}{1-2\tau}$$

$$x(0) = 1.$$

$$\text{Solution: } x(\tau) = \frac{(r-1)(1-2\tau)^{r/2} - (1-2\tau)}{r-2}$$

The smallest positive solution to  $x(\tau) = 0$  is

$$\tau_0 = \frac{1}{2} \left( 1 - \left( \frac{1}{r-1} \right)^{2/(r-2)} \right)$$

and then number of vertices in independent set chosen by GREEDY is whp,  $\approx \tau_0 n$ .

For the following:

$q_0, q_1, \dots, q_t, \dots, q_n \in S$  is a random process.

$H_t = (q_0, q_1, \dots, q_t)$  is the history to time  $t$ .

$X(0), X(1), \dots, X(t), \dots$  are random variables where

$$X(t) = X_t(H_t).$$

$D \subseteq \mathbb{R}^2$  is open and connected and

$$\left(0, \frac{X_0(q_0)}{r_0}\right) \in S$$

[We can assume  
[ $q_0$  is fixed

We further assume

$$(i) \quad |X(t)| \leq C_0 n, \quad \forall t < T_D \text{ where } C_0 \text{ is constant.}$$

$$(ii) \quad |X(t+1) - X(t)| \leq \beta = \beta(n) \geq 1, \quad \forall t < T_D$$

$$(iii) \quad |E(X(t+1) - X(t) | \mathcal{H}_t) - f(t/n, X(t)/n)| \leq \lambda_0, \\ \forall t < T_D$$

(iv)  $f(t, x)$  is **continuous** and satisfies a **Lipschitz** condition on  $D_n \{(t, x) : t \geq 0\}$

$$\text{i.e.} \quad |f(x) - f(x')| \leq L \|x - x'\|_\infty.$$

## Example 1

$$H_B = (i_1, i_2), (i_3, i_4), \dots, (i_{2b-1}, i_{2b}) \quad 1 \leq i_k \leq n$$

$$X_B(H_B) = n - |\{i_1, i_2, \dots, i_{2b}\}| \quad C_0 = 1$$

$$f(B, x) = -1 - x$$

$$\lambda_0 = \frac{1}{n-1}$$

$$L = 1$$

$$D = (-1, 1)^2$$

## Example 2

$$W = [rn]$$

$$H_v = (i_1, i_2), (i_3, i_4), \dots, (i_{2v-1}, i_{2v}) \quad 1 \leq i_k \leq rn$$

$$X_v(H_v) = n - |\{a : \exists s \text{ s.t. } i_s \in W_a\}|$$

$$f(v, \alpha) = -1 - \frac{r\alpha}{1-2v}$$

$$C_0 = 1$$

$$\lambda_0 = \frac{1}{2}n$$

$$L = \frac{r}{2}\alpha$$

$$D = (-1, \frac{1}{2} - \alpha) \times (0, 1)$$

# Theorem

Suppose  $\lambda > \lambda_0$  and  $C$  is sufficiently large and

$\sigma = \inf \{ \tau : (\tau, z(\tau)) \notin D_0 = \{ (t, z) \in D : \text{distance of } (t, z) \text{ to boundary of } D \geq C\lambda \} \}$

where  $z(t)$ ,  $0 \leq \tau \leq \sigma$  be the unique solution to

$$\dot{z}(\tau) = f(\tau, z) \quad (*)$$

$$z(0) = \frac{x_0(\tau_0)}{n}$$

With probability  $1 - O\left(\frac{\beta}{\lambda} \exp\left(-\frac{n\lambda^3}{\beta^3}\right)\right)$

$$X(t) = n z(t/n) + O(\lambda n)$$

uniformly in  $0 \leq t \leq \sigma n$ .

## Proof

$$\text{Let } \omega = \left\lceil \frac{n\lambda}{\beta} \right\rceil.$$

We can assume that  $\frac{\lambda}{\beta} \geq n^{-1/3}$  else there is nothing to prove.

We study the concentration of  $X(t+\omega) - X(t)$ ,

so assume that  $(t/n, X(t/n)) \in D_0$ .

For  $0 \leq k \leq \omega$  we have

$$\text{Note that } \left| \frac{X(t+k)}{n} - \frac{X(t)}{n} \right| \leq \frac{k\beta}{n} \leq 2\lambda$$

$$\text{So } \left\| \left( \frac{t+k}{n}, \frac{X(t+k)}{n} \right) - \left( \frac{t}{n}, \frac{X(t)}{n} \right) \right\|_{\infty} \leq 2\lambda$$

and so  $\left\{ \left( \frac{t+k}{n}, \frac{X(t+k)}{n} \right) \right\}$  is in  $D$ , assuming  $C \geq 2_0$ .

$$\mathbb{E} ( X(t+k+1) - X(t+k) \mid H_{t+k} ) =$$

$$f\left(\frac{t+k}{n}, \frac{X(t+k)}{n}\right) + \theta_k = \quad |\theta_k| \leq \lambda$$

$$f\left(\frac{t}{n}, \frac{X(t)}{n}\right) + \psi_k + \theta_k = \quad |\psi_k| \leq \frac{L\beta k}{n}$$

$$f\left(\frac{t}{n}, \frac{X(t)}{n}\right) + \rho$$

where  $|\rho| \leq 2L\lambda$ .



Now, given  $H_t$ , let

$$Z_k = X(t+k) - X(t) - k f\left(\frac{t}{n}, \frac{X(t)}{n}\right) - 2kL\lambda.$$

Then

$$E(Z_k - Z_{k-1} | Z_0, \dots, Z_{k-1}) \leq 0$$

i.e.  $Z_0, Z_1, \dots, Z_w$  is a **supermartingale**.

Also

$$|Z_k - Z_{k-1}| \leq \beta + \left| f\left(\frac{t}{n}, \frac{X(t)}{n}\right) \right| + 2L\lambda$$

$$\leq K\beta$$

where  $K_0 = O(1)$ .

$O(1)$  by continuity and boundedness of  $S$ .

So, conditional on  $H_F$ ,

$$\begin{aligned} \Pr(X(t+\omega) - X(t) - \omega f(t/n, X(t/n)) \geq 2L\omega\lambda + K_0\beta\sqrt{2\alpha\omega}) \\ \leq e^{-\alpha}. \end{aligned}$$

Similarly,

$$\begin{aligned} \Pr(X(t+\omega) - X(t) - \omega f(t/n, X(t/n)) \leq -2L\omega\lambda - K_0\beta\sqrt{2\alpha\omega}) \\ \leq e^{-\alpha}. \end{aligned}$$

Here we produce a **supermartingale** or equivalently consider  $-X(t)$ .

Thus

$$\Pr(|X(t+w) - X(t) - w f(t/n, X(t/n))| \geq \overbrace{2Lw\lambda + K_0\beta\sqrt{2\alpha w}}^{\text{err}}) \leq 2e^{-\alpha}.$$

We will choose

$$\alpha = \frac{n\lambda^3}{\beta^3}$$

so that  $w\lambda$  and  $\beta\sqrt{2\alpha w}$  are both  $\Theta(n\lambda^2/\beta)$  giving

$$\text{err} \leq K_1 \frac{n\lambda^3}{\beta}$$

Now let  $k_j = jw$  for  $j = 0, 1, \dots, j_0 = \lfloor \sigma n/w \rfloor$ .

We will show by induction that

$$P_j(\exists i \leq j : |X(k_j) - z(k_j/n)| \geq B_j) \leq 2j e^{-\alpha}$$

where

$$B_j = B \left( \left( 1 + \frac{Lw}{n} \right)^j - 1 \right) \frac{n\lambda^2}{\beta}$$

and where  $B$  is another constant.

The induction begins with  $z(0) = \frac{X(0)}{n}$ .

Note that  $B_{j_0} = O\left(\frac{n\lambda^2}{\beta}\right) = O(\lambda n)$ .

Now write

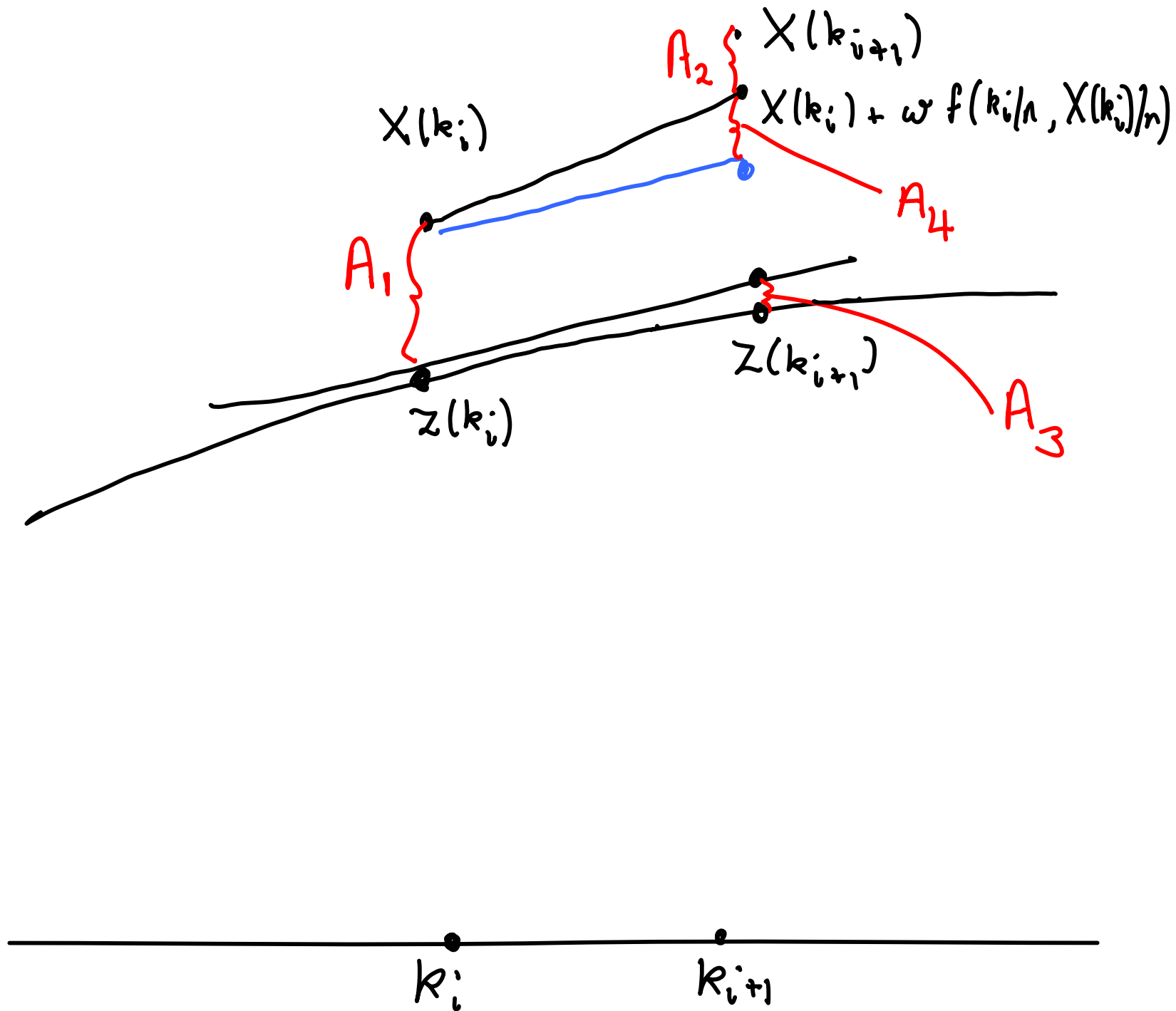
$$|X(k_{i+1}) - z(k_{i+1}/n)n| = |A_1 + A_2 + A_3 + A_4|$$

$$A_1 = X(k_i) - z(k_i/n)n$$

$$A_2 = \cancel{X(k_{i+1})} - X(k_i) - \omega f(k_i/n, X(k_i/n))$$

$$A_3 = \omega z'(k_i/n) + z(k_i/n)n - z(k_{i+1}/n)n$$

$$A_4 = \omega f(k_i/n, X(k_i/n)) - \omega z'(k_i/n)$$



$$A_1 = X(k_i) - z(k_i/n) n$$

The induction gives

$$\|A_1\| \approx B_i.$$

$$A_2 = X(k_{i+1}) - X(k_i) - \omega f(k_i/n, X(k_i/n))$$

$$|A_2| \leq K_1 \frac{n\lambda^2}{\beta}$$

with probability  $1 - 2e^{-\alpha}$ .

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$$A_3 = \omega z'(k_i/n) + z(k_i/n)n - z(k_{i+1}/n)n$$

$$|A_3| \leq L \frac{\omega^2}{n^2} \cdot n = L \frac{\omega^2}{n} \leq 2Ln \frac{\lambda^2}{\beta^2}$$



$$A_4 = \omega f(k_i/n, X(k_i)/n) - \omega Z'(k_i/n)$$

$$|A_4| \leq \frac{\omega L A_1}{n} \leq \frac{\omega L}{n} B_i.$$

Thus, for some  $B > 0$ ,

$$B_{i+1} \leq |A_1| + |A_2| + |A_3| + |A_4|$$

$$\leq \left(1 + \frac{\omega L}{n}\right) B_i + B n^{\frac{\lambda^2}{\beta}}.$$

Finally consider  $k_i \leq t < k_{i+1}$ .

From "time"  $k_i$  to  $t$ , the change in  $X$  and  $nZ$  is at most  $\omega\beta = O(n\lambda)$ .



The above proof generalises easily to the case where

(i)  $X(t)$  is replaced by  $X_1(t), X_2(t), \dots, X_a(t)$  where  $a = O(1)$ .

(ii) Condition (iii) on P11 holds with probability  $1 - \gamma$ .

This adds  $O(n\gamma)$  to the error probability.

We simply condition on (iii) always holding.