Summary: Lecture 10

Summary for the chapters 11.1 up to page 245 and 11.2 (page 258 optional). [5, 1]

Randomized algorithms

Algorithms based on randomization.

(The algorithm employs a degree of randomness as part of its logic or procedure.)

Bipartite matching

Bipartite Graph

A graph G = (U, V, E) is called bipartite if the vertices can be divided into two disjoint and independent sets U and V. (There are no edges between two elements of U or two elements of V).

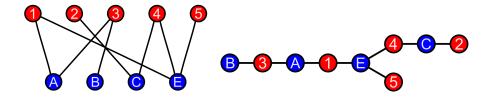


Figure 1: Examples of bipartife graphs with U and V marked in red and blue [2]

Problem: BipartiteMatching

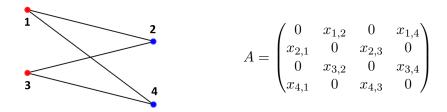
Given: Bipartite graph G = (U, V, E).

Is there a perfect matching $M \subseteq E$ such that for any two edges (u, v) and (u', v') in M $u \neq u'$ and $v \neq v'$.

I other words: A matching in a Bipartite Graph is a set of the edges chosen in such a way that no two edges share an endpoint. The matching M is called perfect if for every node in V there is some edge in M. [3, 5, 4]

- construct bipartite graph with n nodes as $n \times n$ matrix A
- the element $A_{i,j}$ is a variable $x_{i,j}$ if $(i,j) \in E$
- the element $A_{i,j}$ is 0 if $(i,j) \notin E$

Example:



Questions:

Is EVERY node cotained in the subset M when it is a perfect matching? Does a perfect matching then only exist with an even number of vertices and |U| = |V|?

Determinant calculation

Leibniz-formula:

$$A = \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} \qquad |A| = a \cdot e \cdot i + b \cdot f \cdot g + c \cdot d \cdot h - g \cdot e \cdot c - h \cdot f \cdot a - i \cdot d \cdot b$$

$$B = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

$$|B| = 1 \cdot 5 \cdot 9 + 2 \cdot 6 \cdot 7 + 3 \cdot 4 \cdot 8 - 7 \cdot 5 \cdot 3 - 8 \cdot 6 \cdot 1 - 9 \cdot 4 \cdot 2 = 0$$

$$|A| = \sum_{\pi} \sigma(\pi) \prod_{i=1}^{n} A_{i,\pi(i)}$$

- $\sigma(\pi)$ decides if + or -
- leads to n! summands
- Example: n = 3 3! = 6 summands 6 permutations for π

+1:
$$(123) (231) (312)$$

-1: $(321) (213) (132)$
 $\rightarrow |A| = A_{1,1} \cdot A_{2,2} \cdot A_{3,3} + A_{2,1} \cdot A_{3,2} \cdot A_{1,3} + ...$

Gaussian elimination:

- Gauß algorithm for solving LSE (linear systems of equations)
- allowed operations:
 - addition of rwos
 - subtraction of rows
 - multiply row with integer x
 - divide row by integer x
 - switch to rows
- wanted: upper triangular form (all entries below the diagonal 0)
- determinant is the product of the diagonal entries
- Example:

$$\begin{pmatrix} 1 & 3 & 2 & 5 \\ 1 & 7 & -2 & 4 \\ -1 & -3 & -2 & 2 \\ 0 & 1 & 6 & 2 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 3 & 2 & 5 \\ 0 & 4 & -4 & -1 \\ 0 & 0 & 0 & 7 \\ 0 & 1 & 6 & 2 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 3 & 2 & 5 \\ 0 & 4 & -4 & -1 \\ 0 & 0 & 0 & 7 \\ 0 & 0 & 7 & 2\frac{1}{4} \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 3 & 2 & 5 \\ 0 & 4 & -4 & -1 \\ 0 & 0 & 7 & 2\frac{1}{4} \\ 0 & 0 & 0 & 7 \end{pmatrix}$$

Figure 2: Examples gaussian elimination [5]

Symbolic Determinants

Symbolic matrix:

- matrix with variables instead of numerical entries
- Example:

$$\begin{pmatrix} x & w & z \\ z & x & w \\ y & z & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} x & w & z \\ 0 & \frac{x^2 - zw}{x} & \frac{wx - z^2}{x} \\ 0 & \frac{zx - wy}{x} & -\frac{zy}{x} \end{pmatrix} \Rightarrow$$

$$\begin{pmatrix} x & w & z \\ 0 & \frac{x^2 - zw}{x} & \frac{wx - z^2}{x} \\ 0 & 0 & -\frac{yz(xz - xw) + (zx - wy)(wx - z^2)}{x(x^2 - zw)} \end{pmatrix}$$

Figure 3: Examples gaussian elimination with symbolic matrix [5]

- subdeterminants have exponentially many terms
- wanted result: know if determinant is 0

Monte Carlo algorithm

Check if determinant of symbolic matrix is 0:

- use arbitrary integers for the variables
 → numerical matrix
- calculate determinant of numerical matrix:
 - if not 0:
 determinant of symbolic matrix is not 0
 - if 0:
 determinant of symbolic matrix is probably 0
 - * numbers could be chosen such that the numrical determinant is 0 even though the symbolic one is not 0

Monte Carlo algorithm

Randomized algorithm for deciding if a graph G has a perfect matching with calculating the determinant of the corresponding matrix A to G.

- choose m random integers $i_1, ..., i_m$ between 0 and 2m
- compute the determinant $|A|(i_1,...,i_m)$ with the Gaussian elimnation
- if $|A|(i_1,...,i_m) \neq 0$ reply G has a perfect matching
- if $|A|(i_1,...,i_m)=0$ reply G has probably no perfect matching
- if perfect matching found: decision is reliable and final
- if perfect matching not found: possibility of false negative

Monte Carlo algorithm

(Algorithm above) decides whether a symbolic matrix is **not** indentically to zero.

Reducing chance of false negatives:

- perform many independent experiments
- chose each time random integers (independently)
- \bullet repeat k times the evaluation of the determinant of the symbolic matrix
 - answer always zero: chance that G hs no perfect matching is higher $(1-(\frac{1}{2})^k)$
 - answer different from zero once: perfect matching exists

Monte Carlo algorithm:

- Monte Carlo algorithm has no false positives
- probability of false negatives is bounded
- ullet time needed always polynomical

TODO

Questions:

Randomized complexity classes

TODO

Questions:

- some slides about stuff that is not in the book
- Satz von Rice

TODO

Questions:

References

- [1] Martin Berglund. Lecture notes in Computational Complexity.
- [2] Bipartite graph image source. https://en.wikipedia.org/wiki/Bipartite_graph.
- [3] GeeksforGeeks. Maximum Bipartite Matching. https://www.geeksforgeeks.org/maximum-bipartite-matching/, last opened: 09.12.22.
- [4] Swastik Kopparty. Bipartite Graphs and Matchings. https://sites.math.rutgers.edu/~sk1233/courses/graphtheory-F11/matching.pdf, last opened 09.12.22. 2011.
- [5] Christos H. Papadimitriou. *Computational Complexity*. Addison-Wesley Publishing Company, 1994.