# Finding Perfect Matchings and <br> Completing Latin Rectangles 

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## Outline

- Definitions


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- The Problem


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- The degree of a vertex in a graph is the number of edges that vertex is connected to


## Bipartite Graph

Definition: A Bipartite Graph is a graph such that every vertex can be put into one of two groups, $V_{1}$ and $V_{2}$, with the property that all edges connect vertices in $V_{1}$ to vertices in $V_{2}$ (no edges go between vertices in the same group)

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- We shall label the vertices in $V_{1}$ the numbers $1^{\prime}, 2^{\prime}, \ldots,(n-1)^{\prime}$, n'
- And label the vertices in $V_{2}$ the numbers $1,2, \ldots,(n-1), n$


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- This is notably not a perfect matching, and we cannot easily make it one because there does not exist an edge in our initial graph from 2' to 1



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| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| 1 | 1 | 1 | 1 |
| 2 | 2 | 2 | 2 |
| 3 | 3 | 3 | 3 |
| 4 | 4 | 4 | 4 |

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- Each column is headed by a vertex in $V_{1}$


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- Each column is headed by a vertex in $V_{1}$
- Each other number in the column is a vertex in $V_{2}$ such that there is an edge between the head of a column and each element in its column


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For edges that we removed from the $d=n$ graph, we put an $X$ in the array to indicate that there is not an edge

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- When we pick a vertex of $V_{2}$ to draw an edge to a specific vertex in $V_{1}$, we cannot not pick another edge that shares either of these two vertices
- On the table this corresponds to picking an element, E , in a column, $\mathrm{C}^{\prime}$, and then not picking another element E in a different column and likewise not pick another element from column C'


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To best understand this process, let us find a perfect matching of the previous example:


| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ | 1 | 1 |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
| $X_{3}$ | 3 | $X_{3}$ | 3 |
| 4 | 4 | $X_{4}$ | $X_{4}$ |

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As $K_{c}=2$ for all the columns we can pick a random element from any of the columns, let us then arbitrarily pick column 2' element 3

## The Process

Thus having picked $\left(2^{\prime}, 3\right)$ we shall make all other elements in column 2' and row 3 X :

| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ | 1 | 1 |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
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| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ | 1 | 1 |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
| $X_{3}$ | 3 | $X_{3}$ | $X_{3}$ |
| 4 | $X_{4}$ | $X_{4}$ | $X_{4}$ |

## The Process

There are now three rows left. $K_{1}=K_{3}=2$ while $K_{4}=1$ thus we must pick from column 4 ' the element 1

| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ | 1 | 1 |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
| $X_{3}$ | 3 | $X_{3}$ | $X_{3}$ |
| 4 | $X_{4}$ | $X_{4}$ | $X_{4}$ |

## The Process

Having picked $\left(4^{\prime}, 1\right)$ we must now $X$ out all other elements in Column 4' and row 1 :

| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ |  | 1 |
| 2 | 1 |  |  |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
| $X_{3}$ | 3 | $X_{3}$ | $X_{3}$ |
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| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ |  | $X_{1}$ |
| 2 | 1 |  |  |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
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## The Process

We now have that $K_{1}=2$ and $K_{3}=1$ so we must pick from column 3' element 2 :

| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ | $X_{1}$ | 1 |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
| $X_{3}$ | 3 | $X_{3}$ | $X_{3}$ |
| 4 | $X_{4}$ | $X_{4}$ | $X_{4}$ |

## The Process

Putting $X$ 's in the appropriate places yields:

| $1^{\prime}$ | $2^{\prime}$ | $3^{\prime}$ | $4^{\prime}$ |
| :---: | :---: | :---: | :---: |
| $X_{1}$ | $X_{1}$ | $X_{1}$ | 1 |
| 2 | $X_{2}$ | 2 | $X_{2}$ |
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| $X_{1}$ | $X_{1}$ | $X_{1}$ | 1 |
| $X_{2}$ | $X_{2}$ | 2 | $X_{2}$ |
| $X_{3}$ | 3 | $X_{3}$ | $X_{3}$ |
| 4 | $X_{4}$ | $X_{4}$ | $X_{4}$ |

Leaving only element $\left(1^{\prime}, 4\right)$ to pick. The edges defined by $\left(1^{\prime}, 4\right)$, $\left(2^{\prime}, 3\right),\left(3^{\prime}, 2\right)$, and $\left(4^{\prime}, 1\right)$ thus form a perfect matching:


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- If we look at the step in the process before $K_{A_{0}}=0$ then column $A_{0}$ must have had an element, $e_{0}$
- Because we must pick from the column with the fewest elements to pick from, the only reason we would not have select $\left(A_{0}^{\prime}, e_{0}\right)$ would be if there were another column, $A_{1}^{\prime}$, that also only had one element, $e_{1}$, left to pick


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- If we look at the step in the process before $K_{A_{0}}=0$ then column $A_{0}$ must have had an element, $e_{0}$
- Because we must pick from the column with the fewest elements to pick from, the only reason we would not have select $\left(A_{0}^{\prime}, e_{0}\right)$ would be if there were another column, $A_{1}^{\prime}$, that also only had one element, $e_{1}$, left to pick
- And specifically the only reason ( $A_{1}^{\prime}, e_{1}$ ) would prevent ( $A_{0}^{\prime}, e_{0}$ ) from getting chosen is if $e_{1}=e_{0}$ so as to then change $e_{0}$ to an $X$


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- If we now look at the pick that lead to $K_{A_{0}}=K_{A_{1}}=1$ then the columns $A_{0}^{\prime}$ and $A_{1}^{\prime}$ must've had a second element in them.


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- We get that the only reason we would not have selected either $\left(A_{0}^{\prime}, a_{0}\right)$ or $\left(A_{1}^{\prime}, a_{1}\right)$ is if there was a third column, $A_{2}^{\prime}$ that had also had at most two elements left.


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- For $A_{2}^{\prime}$ to be selected and leave behind the situation we had above (going one step back from $K_{A_{1}}=0$ ) we get that all three columns must've shared an element that was selected in column $A_{2}^{\prime}$, and we shall call this element $e_{2}$


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- If we continue this process going P picks back from $K_{A_{1}}=0$ we are considering $\mathrm{P}+1$ columns, $A_{0}, A_{1}, \ldots A_{p}$, each of which has P elements to choose from


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- We also get that these $\mathrm{P}+1$ rows must share at least one element in common so as to lead to the scenario of $\mathrm{P}-1$ picks away from $K_{A_{1}}=0$


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- If we let $\mathrm{P}=\mathrm{n}-\mathrm{d}$ then we arrive at the fact that there are $(n-d)+1$ rows that have at least 1 element, $E$, in common from which to choose


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- If we let $\mathrm{P}=\mathrm{n}$ - d then we arrive at the fact that there are $(n-d)+1$ rows that have at least 1 element, $E$, in common from which to choose
- However we have only removed d edges from our the base graph $\mathrm{d}=\mathrm{n}$


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- However we have only removed d edges from our the base graph $\mathrm{d}=\mathrm{n}$ which means that each element is removed d times across all of the columns, thus each row has an element appear n-d times, but we now have a row, E, can be chosen by $\mathrm{n}-\mathrm{d}+1$ columns, thus


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- However we have only removed d edges from our the base graph $\mathrm{d}=\mathrm{n}$ which means that each element is removed d times across all of the columns, thus each row has an element appear n-d times, but we now have a row, E, can be chosen by $n-d+1$ columns, thus WE HAVE A CONTRADICTION!
- This means that the process detailed in this presentation will always yield a perfect matching


## Applications

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| 2 | 4 | 5 | 1 | 3 |
| 3 | 5 | 4 | 2 | 1 |
| 4 | 3 | 1 | 5 | 2 |
| 5 | 1 | 2 | 3 | 4 |

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Definition: A Latin Rectangle is a partial Latin Square that has the first D rows given and the rest blank
Example of a 2 by 4 Latin Rectangle:

| 1 | 2 | 3 | 4 |
| :--- | :--- | :--- | :--- |
| 2 | 4 | 5 | 1 |
|  |  |  |  |
|  |  |  |  |

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| Bipartite Graph <br> Features | Analogous Features of <br> Latin Squares |
| :---: | :---: |
| Each vertex in $V_{1}$ | One specific Column |

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| Bipartite Graph <br> Features | Analogous Features of <br> Latin Squares |
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As we can create d perfect matchings of a Bipartite Graph, we can complete the Latin Rectangle

## Thank you

Thank You all for listening to this presentation Questions?

