## What is...Bondy's theorem?

Or: Forgetting without loss

## From cubes to squares



- Projecting the three points to the indicated planes keeps them unequal
- One checks that this is can be achieved for any three points
- Question Is something going on?
$\left[\begin{array}{lll}1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 1\end{array}\right)$
- Removing the indicated column/row keeps the rows/columns unequal
- The same works for any $4 \times 4$ matrix with $0-1$ entries
- Question Is something going on?

Three points in $\mathbb{R}^{3}$

- Projecting the three points to the xy plane keeps them distinct
- Changing the projection plane, the same works for any three points
- Question Is something going on?


## Enter, the theorem

Let $S$ be a set with $n$ elements and suppose that $n$ distinct subsets of $S$ are chosen.
Then there is a restriction to $n-1$ elements of $S$ under which these subsets remain distinct

## How could that be true?

- Think of $n$ distinct vectors in $\mathbb{R}^{n}$
- Forgetting the $i$ th coordinate identifies two $\Rightarrow$ the line between them is parallel to $e_{i}$
- No projection distinguishes them $\Rightarrow$ there are $n$ pairwise orthogonal lines connecting them
- $n$ points lie inside an $(n-1)$ dim subspace, e.g. three points in $\mathbb{R}^{3}$ :

- Hmm, $n-1$ and $n$ doesn't want to go along

Generalizing Bondy's theorem


- A version of Bondy's theorem Given a collection of distinct vertices on the $n$ cube, what is the largest $d$ such that some projection on $n-d$ dimensions results in a $d$ cube?
- Above $d=2$
- This VC dimension $d$ is an important concept in machine learning

Thank you for your attention!

I hope that was of some help.

