

# Math 201

Section F03

September 29, 2021

# Colloquium

Remember to advertise the colloquium!

## Range and nullspace

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**Remark.** If  $f: V \rightarrow W$  is a linear function, and  $B$  is a basis for  $V$ , then

$$\text{im}(f) = \text{Span}(f(B)).$$

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It is a subspace of  $V$  (by Proposition 2). The dimension of the kernel is called the *nullity* of  $f$  (provided it is finite-dimensional) and is denoted  $\text{nullity}(f)$ .

## Rank-nullity theorem

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## Rank-nullity theorem

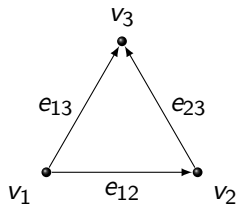
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In other words,

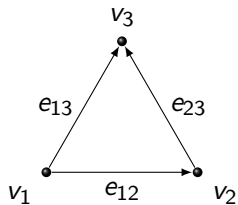
$$\dim(\text{im}(f)) + \dim(\ker(f)) = \dim V.$$

## A little graph theory



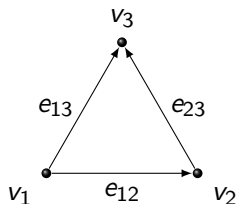


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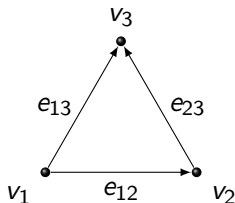


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$$e_{ij} \mapsto v_j - v_i.$$

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**Question:** What is the kernel of  $\partial$ ?