

# Advanced Mathematics 2 - Linear Algebra

## Chapter 5: The vector space $R^n$

Department of Mathematics  
The FPT university

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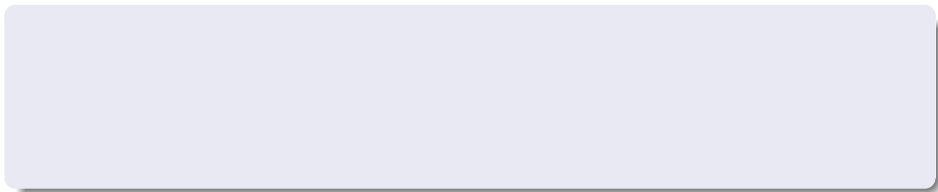
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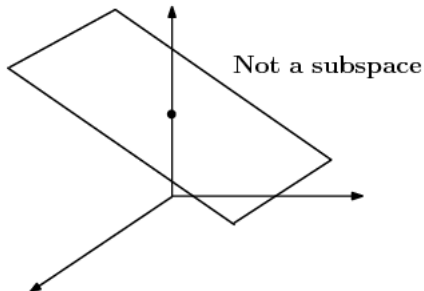
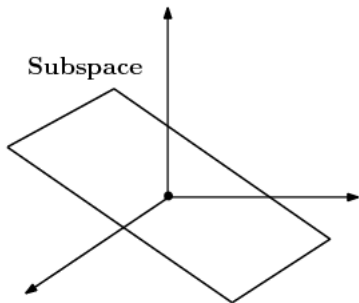




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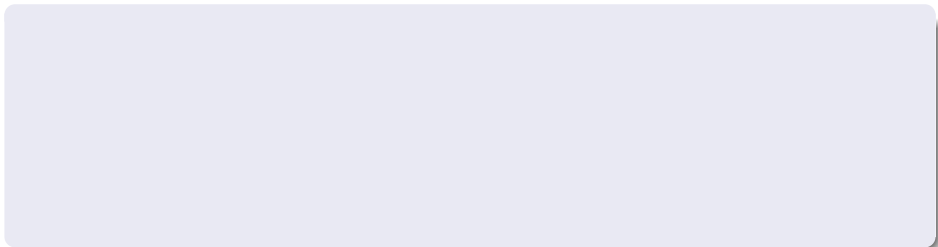
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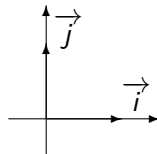
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- The set  $\{\vec{X}_1, \vec{X}_2, \dots, \vec{X}_k\}$  is linearly independent if and only if the system

$$\begin{bmatrix} \vec{X}_1 & \vec{X}_2 & \dots & \vec{X}_k \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ \vdots \\ t_k \end{bmatrix} = 0$$

has only the trivial solution  $t_1 = t_2 = \dots = t_k = 0$ .





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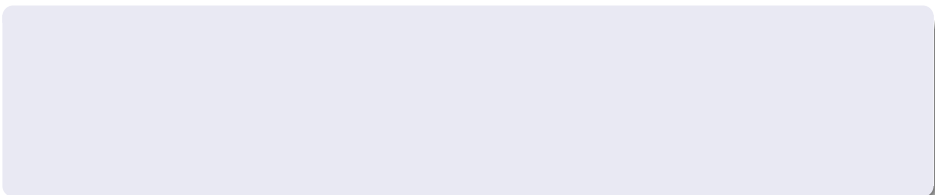
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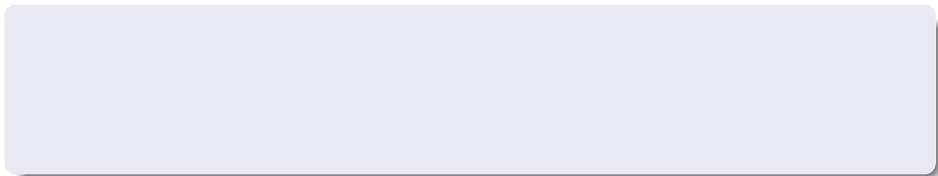
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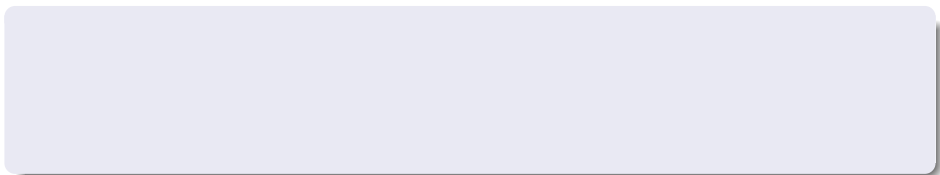
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This is equivalent to determine a maximum set of independent vectors in the set  $\{\vec{X}_1, \vec{X}_2, \dots, \vec{X}_m\}$



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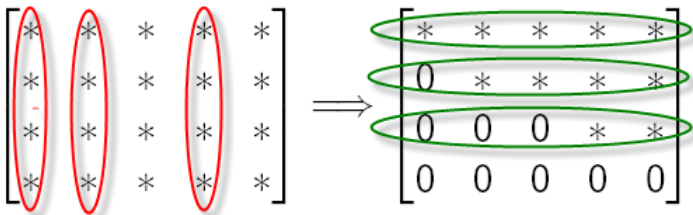
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