

## Smoothness of the Function and Laplacian

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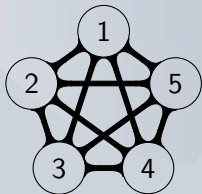
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The smoothness of  $k$ -th eigenvector is the  $k$ -th eigenvalue.



# Laplacian of the Complete Graph $K_N$

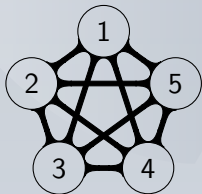
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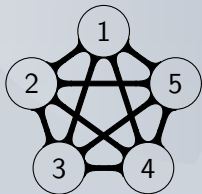
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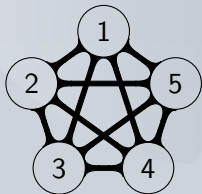
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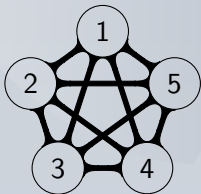
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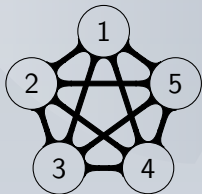
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Answer:  $N-1$  eigenvectors  $\perp \mathbf{1}_N$  for eigenvalue  $N$  with multiplicity  $N-1$

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$(\lambda, \mathbf{u})$  is an eigenpair for  $\mathbf{L}_{rw}$  iff  $(\lambda, \mathbf{D}^{1/2} \mathbf{u})$  is an eigenpair for  $\mathbf{L}_{sym}$



# Michal Valko

`michal.valko@inria.fr`

Inria & ENS Paris-Saclay, MVA

`https://misovalko.github.io/mva-ml-graphs.html`