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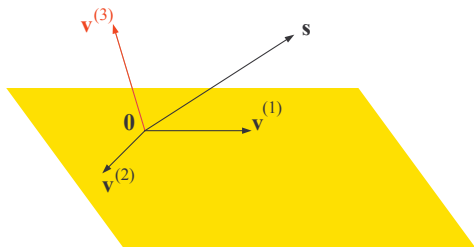
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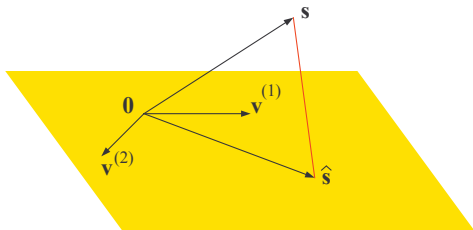
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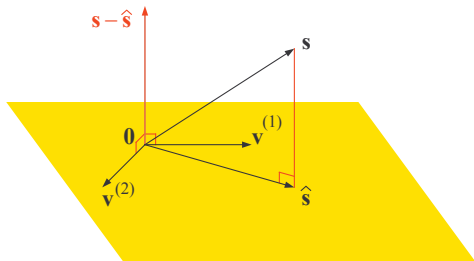
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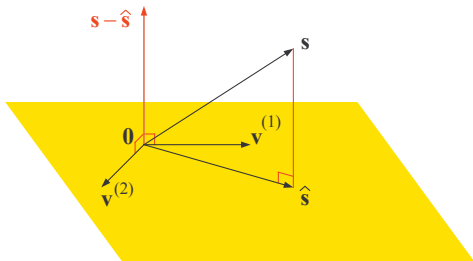
$\mathbf{V}\mathbf{c}$  can **no longer** represent every  $\mathbf{s} \in \mathbb{R}^m$ .



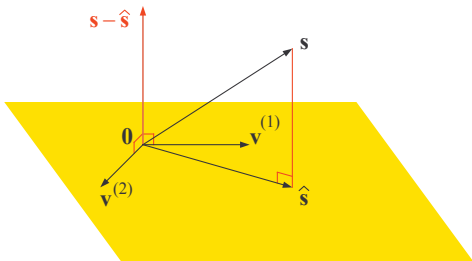




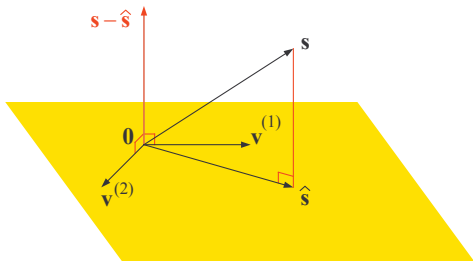




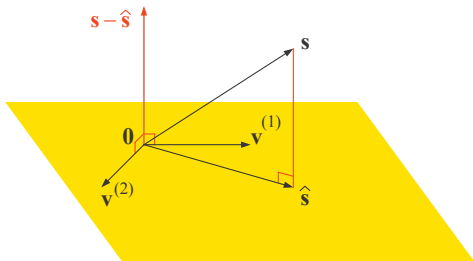
Error norm  $\|s - \hat{s}\|$



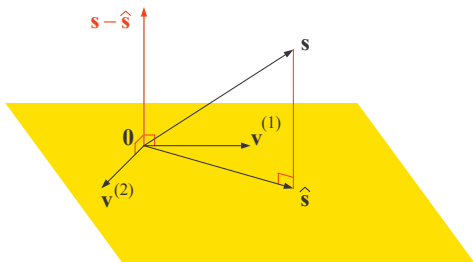
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$$\mathbf{s} - \hat{\mathbf{s}} \perp \mathbf{v}^{(1)}, \dots, \mathbf{v}^{(n)}$$

## The Orthogonal case



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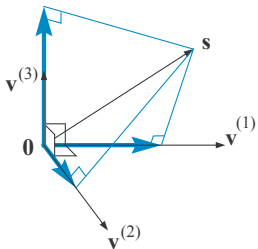
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