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Figure 1: Decomposition $\mathbf{J} = \mathbf{U}_J \mathbf{\Sigma}_J \mathbf{V}_J^\top$ in the case where $r_J = 2$. Each of the joint components has an n -dimensional score vector (column of \mathbf{U}_J) and a p -dimensional loading vector (row of \mathbf{V}_J^\top) associated with it; in this example, $n = 3$ and $p = 4$. Each subward has an r_J -dimensional score vector (row of \mathbf{U}_J) associated with it, and each feature has a r_J -dimensional loading vector (column of \mathbf{V}_J^\top). Given that the singular values in $\mathbf{\Sigma}_J$ are distinct and ordered from largest to smallest, the decomposition is identifiable up to multiplication of components by -1: we can multiply any column of \mathbf{U}_J and the corresponding column of \mathbf{V}_J (row of \mathbf{V}_J^\top) by -1 without changing the value of $\mathbf{U}_J \mathbf{\Sigma}_J \mathbf{V}_J^\top$.