A rank one matrix is a matrix of the form uv^T for column vectors u and v. This is the opposite order of the dot product, $u^Tv = \langle u, v \rangle$. If u has m rows, and v has n rows then uv^T is a $m \times n$ matrix with m rows and n columns.

Today we only care about the case when u = v, where something magical happens!

Let
$$u = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$
. Find the matrix $\begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 2 \end{bmatrix}$. $\begin{bmatrix} 1 & 0 & 2 \end{bmatrix}$.

What are its eigenpairs? Use Ch5
$$\left(5, \begin{bmatrix} 2 \\ 2 \end{bmatrix}\right) \left(0, \begin{bmatrix} -22 \\ 4 \end{bmatrix}\right)$$
both zero

What do you notice about the matrix?

$$Symmetric (\vec{u}\vec{u}^T)^T = \vec{u}\vec{u}^T$$

What do you notice about its eigenvectors?

you notice about its eigenvectors?
$$\vec{u}\vec{v}$$
 has eigen pairs $(\vec{u}\vec{v},\vec{u})$ and $(0,\vec{v})$ for any nonzero \vec{v} or thogonal to \vec{u}

Symmetric matrices

A matrix A is symmetric iff $A = A^T$. Symmetric matrices have to be square, so they can have eigenvalues and eigenvectors. In fact, the eigenvalues and eigenvectors are very nice: The eigenvalues are all real numbers (no square roots of negatives) and the eigenvectors for distinct eigenvalues are orthogonal (so easy to project onto). If A is $n \times n$, then it has exactly n eigenvalues (counted with multiplicity) and \mathbb{R}^n has an orthogonal basis of eigenvectors.

Symmetric matrices are exactly the $n \times n$ matrices that can be written as a sum

$$c_1\vec{\mathbf{u}}_1\vec{\mathbf{u}}_1^T + c_2\vec{\mathbf{u}}_2\vec{\mathbf{u}}_2^T + c_3\vec{\mathbf{u}}_3\vec{\mathbf{u}}_3^T + \dots c_n\vec{\mathbf{u}}_n\vec{\mathbf{u}}_n^T$$

with c_i real numbers (the eigenvalues) and $\vec{\mathbf{u}}_i$ unit-length orthogonal column vectors with n rows (the eigenvectors). If you use less than n vectors, the remaining ones can be taken to be $\vec{\mathbf{0}}$. If $\vec{\mathbf{u}}_i$ is not unit-length, then the eigenvalue is actually $c_i \vec{\mathbf{u}}_i^T \vec{\mathbf{u}}_i$, the c_i multiplied by the length squared.