The Gram-Schmidt Algorithm.

Suppose that V is an inner product space and that H is a finite dimensional non-trivial subspace. Note that if H has an orthogonal basis $\mathcal{B} = \{\mathbf{v}_1, \ldots, \mathbf{v}_k\}$, then we may compute the coordinates of any $\mathbf{v} \in H$ relative to H as follows: writing $\mathbf{v} = c_1 \mathbf{v}_1 + \cdots + c_k \mathbf{v}_k$, we take the inner product of both sides with \mathbf{v}_j and discover that $\langle \mathbf{v}, \mathbf{v}_j \rangle = c_j \langle \mathbf{v}_j, \mathbf{v}_j \rangle$. Hence the *j*th coordinate of $[\mathbf{v}_i]_{\mathcal{B}}$ is $c_i = \langle \mathbf{v}, \mathbf{v}_i \rangle / ||\mathbf{v}_i||^2$. Thus

$$\mathbf{v} = \operatorname{Proj}_{H}(\mathbf{v}) := \sum_{j=1}^{k} \frac{\langle \mathbf{v}, \mathbf{v}_{j} \rangle}{\|\mathbf{v}_{j}\|^{2}} \mathbf{v}_{j}$$

for all $\mathbf{v} \in H$. Now if $\mathbf{v} \in V$ is a vector outside H, the righthand expression still defines a vector $\operatorname{Proj}_{H}(\mathbf{v}) \in H$, but this cannot possibly equal \mathbf{v} . Instead $\operatorname{Proj}_{H}: V \to H$ is a linear (why? Think about it.) transformation known as the 'orthogonal projection' of Vonto H. As the word 'the' implies, the transformation turns out the same no matter which orthogonal basis we use to define it.

Proposition 0.1. Let $\operatorname{Proj}_H : V \to H$ be as above. Given $\mathbf{v} \in V$, let $\mathbf{v}_H = \operatorname{Proj}_H(\mathbf{v})$ and $\mathbf{v}_{\perp} = \mathbf{v} - \mathbf{v}_{H}$. Then \mathbf{v}_{H} and \mathbf{v}_{\perp} are the unique vectors satisfying $\mathbf{v}_{H} \in H$, $\mathbf{v}_{\perp} \in H^{\perp}$ and $\mathbf{v} = \mathbf{v}_H + \mathbf{v}_\perp$. Moreover, \mathbf{v}_H is the unique vector in H minimizing the distance $\|\mathbf{v} - \mathbf{v}_H\|$.

Proof. By linearity and the definition of Proj_{H} , we have

$$\langle \mathbf{v}_{\perp}, \mathbf{v}_{j} \rangle = \langle \mathbf{v}, \mathbf{v}_{j} \rangle - \sum_{\ell=1}^{K} \frac{\langle \mathbf{v}, \mathbf{v}_{\ell} \rangle \langle \mathbf{v}_{\ell}, \mathbf{v}_{j} \rangle}{\|\mathbf{v}_{\ell}\|^{2}} = \langle \mathbf{v}, \mathbf{v}_{j} \rangle - \frac{\langle \mathbf{v}, \mathbf{v}_{j} \rangle \langle \mathbf{v}_{j}, \mathbf{v}_{j} \rangle}{\|\mathbf{v}_{j}\|^{2}} = 0.$$

So \mathbf{v}_{\perp} is orthogonal to each vector $\mathbf{v}_{j} \in \mathcal{B}$, and it follows that $\mathbf{v}_{\perp} \in H^{\perp}$. Clearly $\mathbf{v} = \mathbf{v}_{H} + \mathbf{v}_{\perp}$. If $\mathbf{w}_H \in H$ and $\tilde{\mathbf{w}}_{\perp} \in H^{\perp}$ also satisfy $\mathbf{v} = \mathbf{w}_H + \mathbf{v}_{\perp}$, then $\mathbf{w}_H + \mathbf{w}_{\perp} = \mathbf{v}_H + \mathbf{v}_{\perp}$ implies that $\mathbf{w}_H - \mathbf{v}_H = \mathbf{v}_\perp - \mathbf{w}_\perp \in H \cap H^\perp$. On the other hand, non-degeneracy of the inner product says that **0** is the only vector orthogonal to itself. I conclude that $\mathbf{w}_H - \mathbf{v}_H = \mathbf{v}_{\perp} - \mathbf{w}_{\perp} = \mathbf{0}$. That is, \mathbf{v}_H and \mathbf{v}_{\perp} give the unique decomposition of \mathbf{v} into vectors in and orthogonal to H.

For the final assertion, I let $\mathbf{w} \in H$ be any vector and estimate

$$\|\mathbf{v} - \mathbf{w}\|^2 = \|\mathbf{v}_{\perp} - (\mathbf{w} - \mathbf{v}_H)\|^2 = \langle \mathbf{v}_{\perp} - (\mathbf{w} - \mathbf{v}_H), \mathbf{v}_{\perp} - (\mathbf{w} - \mathbf{v}_H) \rangle.$$

But $\mathbf{w} - \mathbf{v}_H \in H$ is orthogonal to \mathbf{v}_{\perp} , so when I expand the last expression, two of the four terms vanish leaving me with.

$$\|\mathbf{v} - \mathbf{w}\|^{2} = \langle \mathbf{v}_{\perp}, \mathbf{v}_{\perp} \rangle + \langle \mathbf{w} - \mathbf{v}_{H}, \mathbf{w} - \mathbf{v}_{H} \rangle = \|\mathbf{v}_{\perp}\|^{2} + \|\mathbf{w} - \mathbf{v}_{H}\|^{2} \ge \|\mathbf{v}_{\perp}\|^{2},$$

juality only if $\mathbf{w} = \mathbf{v}_{H}$.

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Of course, the definition of orthogonal projection required that H have at least one orthogonal basis. Our final result guarantees us that this will always be the case when H is finite dimensional.

Theorem 0.2. Let H be a non-trivial finite dimensional subspace of an inner product space V. Then H has an orthogonal basis.

Note that the proof doesn't just show that an orthogonal basis exists; it actually gives a recursive method for constructing it. This is known as the 'Gram-Schmidt algorithm.'

Proof. I work by induction on dim H. In the case dim H = 1, any basis $\{\mathbf{v}_1\}$ for H consists of a single non-zero vector and is therefore also an orthogonal set.

Supposing the theorem is true when dim $H \leq k$, I consider the case dim H = k + 1. Let $\mathcal{B} = \{\mathbf{v}_1, \ldots, \mathbf{v}_{k+1}\}$ be a basis for H and let $W = \text{span}\{\mathbf{v}_1, \ldots, \mathbf{v}_k\}$ be the smaller subspace generated by the first k basis vectors. Then by my inductive hypothesis, there is an orthogonal basis $\{\mathbf{w}_1, \ldots, \mathbf{w}_k\}$ for W. By the Proposition, I have $\mathbf{v}_{k+1} = \text{Proj}_W(\mathbf{v}_{k+1}) + \mathbf{w}_{k+1}$ where $\mathbf{w}_{k+1} \in W^{\perp} \cap H$.

I claim $\mathbf{w}_{k+1} \neq \mathbf{0}$. If not, then $\mathbf{v}_{k+1} = \operatorname{Proj}_W(\mathbf{v}_{k+1}) \in W$ must be a linear combination of $\mathbf{v}_1, \ldots, \mathbf{v}_k$, which contradicts independence of \mathcal{B} . On the other hand, since $\mathbf{w}_1, \ldots, \mathbf{w}_k \in W$ and $\mathbf{w}_{k+1} \in W^{\perp}$, I have $\langle \mathbf{w}_{k+1}, \mathbf{w}_j \rangle = 0$ for all $j \leq k$. Hence $\mathcal{C} := \{\mathbf{w}_1, \ldots, \mathbf{w}_{k+1}\}$ is an orthogonal set of non-zero vectors and therefore independent. Since all $\mathbf{w}_j \in H$ and dim H = k + 1, it follows that \mathcal{C} is an orthogonal basis for H. This completes the inductive step and the proof.