## **Orthogonal Projections**

Last class, we projected a vector **y** onto a line that was the span of a vector  $\mathbf{u} \rightarrow$  a subspace with dimension 1

Today, we will discuss projecting a vector  $\mathbf{y}$  onto a subspace that has a dimension greater than 1

Consider  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4, \mathbf{u}_5\}$ , an orthogonal basis of  $\mathbb{R}^5$  and the vector  $\mathbf{y}$  in  $\mathbb{R}^5$ .

Consider the subspace  $W = \text{span}\{\mathbf{u}_1, \mathbf{u}_2\}$ .

$$\mathbf{y} = c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2 + c_3 \mathbf{u}_3 + c_4 \mathbf{u}_4 + c_5 \mathbf{u}_5$$
  
$$\mathbf{y} = (c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2) + (c_3 \mathbf{u}_3 + c_4 \mathbf{u}_4 + c_5 \mathbf{u}_5)$$
  
$$\mathbf{y} = \mathbf{z}_1 + \mathbf{z}_2$$

$$\Rightarrow \mathbf{z}_1 \text{ is in } W, \text{ let's show } \mathbf{z}_2 \text{ is in } W^{\perp}$$

$$\text{Show } \mathbf{z}_2 \text{ is orthog. to } \mathbf{u}_1 \text{ and } \mathbf{u}_2$$

$$= (c_3 \mathbf{u}_3 + c_4 \mathbf{u}_4 + c_5 \mathbf{u}_5) \cdot \mathbf{u}_1 = c_3(\mathbf{u}_3 \cdot \mathbf{u}_1) + c_4(\mathbf{u}_4 \cdot \mathbf{u}_1) + c_5(\mathbf{u}_5 \cdot \mathbf{u}_1) = 0$$

$$= \frac{1}{\mathbf{z}_2 \cdot \mathbf{u}_1} = (c_3 \mathbf{u}_3 + c_4 \mathbf{u}_4 + c_5 \mathbf{u}_5) \cdot \mathbf{u}_1 = c_3(\mathbf{u}_3 \cdot \mathbf{u}_1) + c_4(\mathbf{u}_4 \cdot \mathbf{u}_1) + c_5(\mathbf{u}_5 \cdot \mathbf{u}_1) = 0$$

$$= \frac{1}{\mathbf{z}_2 \cdot \mathbf{u}_2} = \frac{1}{\mathbf{u}_2} \cdot \mathbf{u}_2 = \frac{$$

This means that  $W^{\perp} = \operatorname{span}{\{\mathbf{u}_3, \mathbf{u}_4, \mathbf{u}_5\}}$ 

Thm. Orthogonal Decomposition Theorem

Let *W* be a subspace of  $\mathbb{R}^n$ . Every vector **y** in  $\mathbb{R}^n$  can be written uniquely in the form

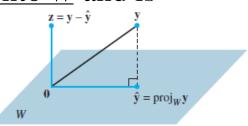
$$\mathbf{y}=\hat{\mathbf{y}}+\mathbf{z}$$

where  $\hat{\mathbf{y}}$  is in *W* and  $\mathbf{z}$  is in  $W^{\perp}$ .

If  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$  is an orthogonal basis of W, then  $\hat{\mathbf{y}} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2 + \dots + \frac{\mathbf{y} \cdot \mathbf{u}_p}{\mathbf{u}_p \cdot \mathbf{u}_p} \mathbf{u}_p$ 

and  $\mathbf{z} = \mathbf{y} - \mathbf{\hat{y}}$ 

 $\hat{\mathbf{y}}$  is called the <u>orthogonal projection of  $\mathbf{y}$  onto W</u> and is sometimes written  $\operatorname{proj}_W \mathbf{y}$ .



$$\underbrace{\operatorname{Ex. Let}}_{Y} \left\{ \begin{bmatrix} 2\\5\\-1\\1\\1\\-1 \end{bmatrix} \right\} \text{ be an orthogonal basis for } W. \text{ Write}$$

$$y = \begin{bmatrix} 1\\2\\3\\-3 \end{bmatrix} \text{ as the sum of a vector in } W \text{ and a vector in } W^{\perp}.$$

$$\widehat{\gamma} = \underbrace{\frac{1}{Y} \cdot \underbrace{u_{1}}_{Y}}_{W_{1} \cdot \underbrace{u_{1}}_{Y}} + \underbrace{\frac{1}{Y} \cdot \underbrace{u_{1}}_{Y}}_{W_{2} \cdot \underbrace{u_{2}}_{Y}} \underbrace{u_{2}}_{Y} = \frac{2 + |0^{-3}}{4 + 25 + 1} \begin{bmatrix} 2\\5\\-1\\-1 \end{bmatrix} + \frac{-2 + 2 + 3}{4 + 1 + 1} \begin{bmatrix} -2\\-2\\-1\\-1 \end{bmatrix} = \begin{bmatrix} 1.6\\-.5\\-.3 \end{bmatrix} + \begin{bmatrix} -1\\-.5\\-.5 \end{bmatrix}$$

$$\widehat{\gamma} = \underbrace{\frac{1}{Y} - \widehat{\gamma}}_{Y} = \begin{bmatrix} 1\\2\\-.2\\-.2 \end{bmatrix} = \begin{bmatrix} 1.4\\0\\2..8 \end{bmatrix} \in W^{\perp} \qquad = \begin{bmatrix} -.4\\2\\.2\\.8 \end{bmatrix}$$

$$\widehat{\gamma} = \widehat{\gamma} - \widehat{\gamma} = \begin{bmatrix} 1\\2\\.2\\.2 \end{bmatrix} + \begin{bmatrix} 1.4\\0\\2..8 \end{bmatrix}$$

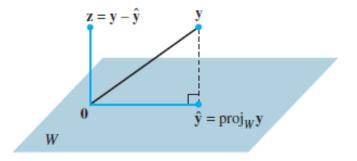
Thm. Best Approximation Theorem

 $\hat{\mathbf{y}}$  is the vector in W that is closest to  $\mathbf{y}$ , in the sense that, for any vector  $\mathbf{v}$  in W,

$$||\mathbf{y} - \hat{\mathbf{y}}|| \le ||\mathbf{y} - \mathbf{v}||$$

 $\hat{\mathbf{y}}$  is called the <u>best approximation of  $\mathbf{y}$  by elements of W.</u>

Because we haven't discussed the basis of W, this means that  $\hat{\mathbf{y}}$  is the same no matter what basis is used for W.



$$\underbrace{Ex. \text{ Find the distance from } \mathbf{y} = \begin{bmatrix} -1\\ -5\\ 10 \end{bmatrix} \text{ to } W = \operatorname{span} \left\{ \begin{bmatrix} 5\\ -2\\ 1 \end{bmatrix}, \begin{bmatrix} 1\\ 2\\ -1 \end{bmatrix} \right\}. \text{ or theg.} \\
\underbrace{\mathbf{y}}_{\mathbf{y}} = \underbrace{\frac{\mathbf{y}}_{\mathbf{y}} \cdot \underbrace{\mathbf{y}}_{\mathbf{y}}}_{\mathbf{y}_{1}} \underbrace{\mathbf{y}}_{\mathbf{y}_{2}} + \underbrace{\frac{\mathbf{y}}_{\mathbf{y}} \cdot \underbrace{\mathbf{y}}_{\mathbf{y}}}_{\mathbf{y}_{2}} \underbrace{\mathbf{y}}_{\mathbf{y}_{2}} = \frac{-5 + 10 + 10}{25 + 14 + 1} \begin{bmatrix} 5\\ -2\\ 1 \end{bmatrix} + \frac{-1 - 40 - 10}{1 + 4 + 1} \begin{bmatrix} 1\\ 2\\ -1 \end{bmatrix} = \begin{bmatrix} 2.5\\ -1\\ 5 \end{bmatrix} + \begin{bmatrix} 3.5\\ 7\\ 3.5 \end{bmatrix} \\
\underbrace{\mathbf{y}}_{\mathbf{y}} = \underbrace{\mathbf{y}}_{\mathbf{y}} \cdot \underbrace{\mathbf{y}}_{\mathbf{y}} = \underbrace{\begin{bmatrix} -1\\ -5\\ 10 \end{bmatrix} - \begin{bmatrix} -1\\ -8\\ 4 \end{bmatrix}} = \begin{bmatrix} 0\\ 3\\ 6 \end{bmatrix} = \begin{bmatrix} 0\\ 3\\ 6 \end{bmatrix} = \begin{bmatrix} -1\\ -8\\ 4 \end{bmatrix} \\
\underbrace{\mathbf{y}}_{\mathbf{y}} = \underbrace{\mathbf{y}}_{\mathbf{y}} \cdot \underbrace{\mathbf{y}}_{\mathbf{y}} = \underbrace{\begin{bmatrix} -1\\ -8\\ 4 \end{bmatrix}} = \begin{bmatrix} 0\\ 3\\ 6 \end{bmatrix} = \begin{bmatrix} -1\\ -8\\ 4 \end{bmatrix} \\
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\underbrace{\mathbf{y}}_{\mathbf{y}} = \underbrace{\begin{bmatrix} -1\\ -8\\ -8\\ -8 \end{bmatrix}} \\
\underbrace{\mathbf{y}}$$

If  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$  is an orthonormal basis of *W*, then  $\hat{\mathbf{y}} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2 + \dots + \frac{\mathbf{y} \cdot \mathbf{u}_p}{\mathbf{u}_p \cdot \mathbf{u}_p} \mathbf{u}_p$ If we define  $U = [\mathbf{u}_1 \ \mathbf{u}_2 \ \cdots \ \mathbf{u}_p]$ , then  $\hat{\mathbf{y}} = UU^T \mathbf{y}$  $\rightarrow$  Prove it  $\widehat{\gamma} = \mathcal{U} \cdot \mathcal{U}^{\mathsf{T}} \widehat{\gamma} = \mathcal{U} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \mathcal{U} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} = \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_{1} \\ \widehat{\mathcal{U}}_{2} \end{bmatrix} \end{bmatrix} \begin{bmatrix} \widehat{\mathcal{U}}_$  $= (\overline{u_1}, \overline{y})\overline{u_1} + (\overline{u_2}, \overline{y})\overline{u_2} + \dots + (\overline{u_p}, \overline{y})\overline{u_p} /$ 

$$\underline{\text{Ex. Let } W = \text{span} \left\{ \begin{bmatrix} 1\\0\\-1 \end{bmatrix}, \begin{bmatrix} 0\\1\\0 \end{bmatrix} \right\} \text{ (note its orthogonal).}$$
Find a matrix A such that  $\text{proj}_{W} \mathbf{y} = A\mathbf{y}$  for any vector  $\mathbf{y}$ .  

$$\vec{w}_{1} = \frac{1}{\sqrt{1+0+1}} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -1 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}$$

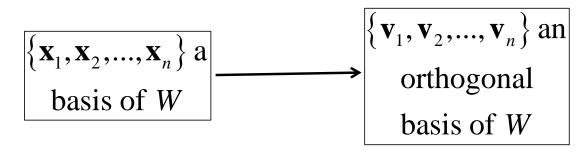
$$\vec{w}_{2} = \begin{bmatrix} 0\\-1\\0 \end{bmatrix}$$

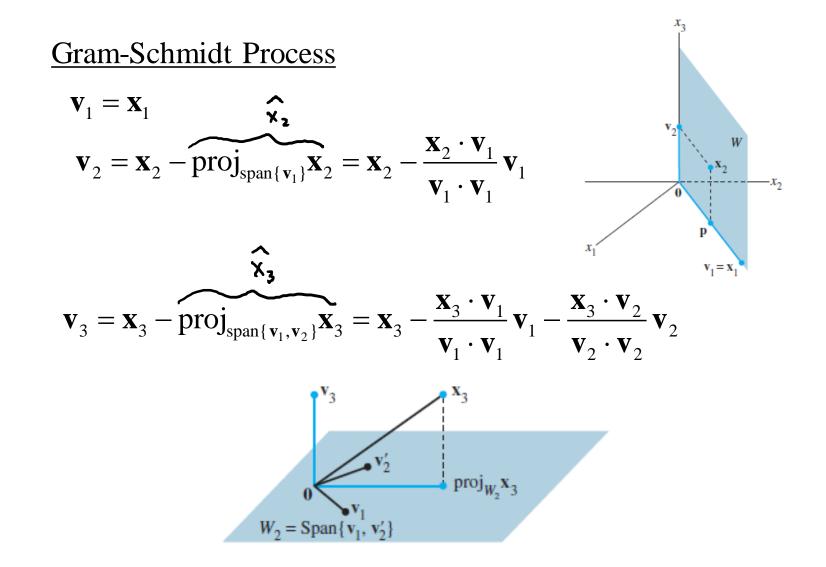
$$\vec{w}_{2} = \begin{bmatrix} 0\\-1\\0 \end{bmatrix}$$

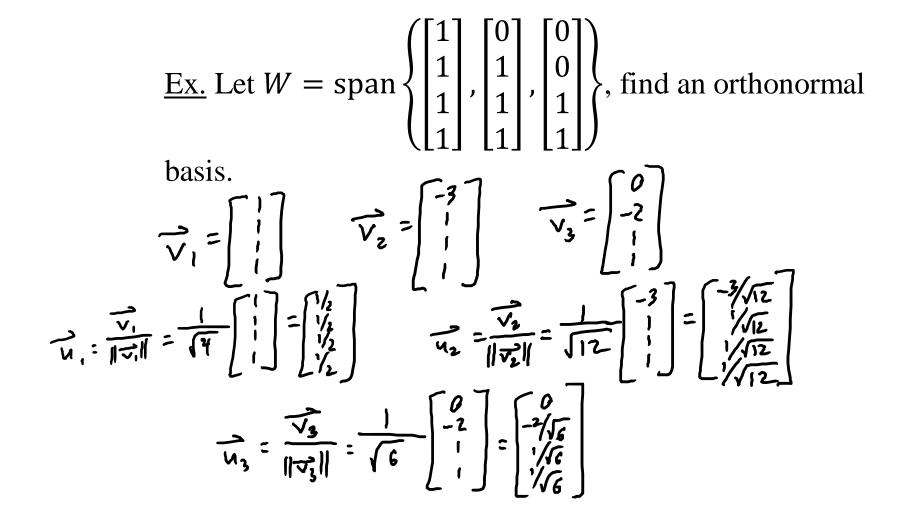
$$\vec{w}_{2} = \begin{bmatrix} 0\\-1\\0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & 0\\0 & 1 & 0\\0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & 0\\0 & 1 & 0\\0 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} & 0\\-\frac{1}{\sqrt{2}} \end{bmatrix}$$

We've seen that it's useful to have an orthogonal basis

 $\rightarrow$  If we are given some other basis, we can find an orthogonal basis using the Gram-Schmidt Process







$$\underbrace{\operatorname{Ex.} \operatorname{Let} W = \operatorname{Nul} \begin{bmatrix} 1 & -1 & -2 \\ 2 & -2 & -4 \\ 4 & -4 & -8 \end{bmatrix}}_{\operatorname{4} \operatorname{4} \operatorname{4} \operatorname{4} \operatorname{-8}} \underbrace{\operatorname{find the point in } W}_{\widehat{V}} \text{ that is}$$

$$\operatorname{closest to } \mathbf{y} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \text{ and find the distance from } \mathbf{y} \text{ to } W.$$

$$\|\mathbf{z}\|$$

$$\underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 2 & -2 & -4 \\ 4 & -4 & -8 \end{bmatrix}}_{V_1} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 2 & -2 & -4 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_1} \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_1} \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{V_2} \Rightarrow \underbrace{\begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ V_3 & V_3 & V_7 & V_7 & V_7 \\ V_1 & V_2 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_2 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 \\ V_1 & V_1 & V_1 & V_1 \\ V_1 & V$$

