

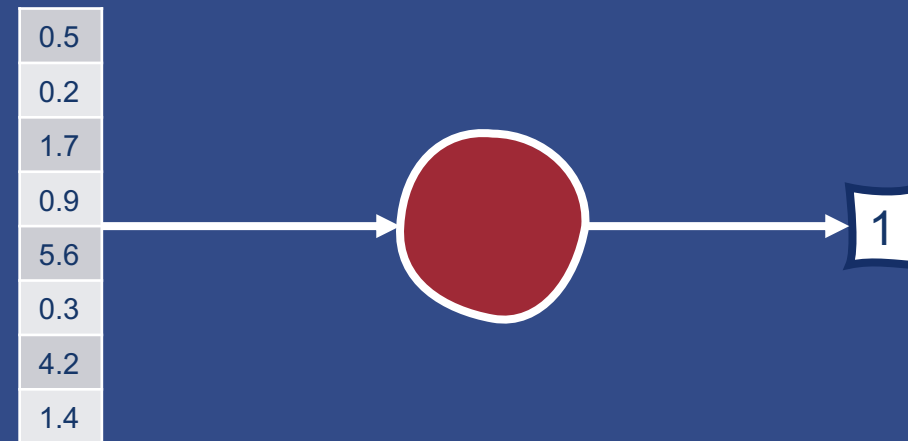
# **Building Blocks for Neural Networks**

Natalie Parde

UIC CS 421

# Building Blocks for Neural Networks

- At their core, neural networks are comprised of **computational units**
- Computational units:
  1. Take a set of real-valued numbers as input
  2. Perform some computation on them
  3. Produce a single output



# Computational Units

- The computation performed by each unit is a weighted sum of inputs
  - Assign a weight to each input
  - Add one additional bias term
- More formally, given a set of inputs  $x_1, \dots, x_n$ , a unit has a set of corresponding weights  $w_1, \dots, w_n$  and a bias  $b$ , so the weighted sum  $z$  can be represented as:
  - $z = b + \sum_i w_i x_i$

# Sound familiar?

- This is exactly the same sort of weighted sum of inputs that we needed to find with logistic regression!
- Recall that we can also represent the weighted sum  $z$  using vector notation:
  - $z = w \cdot x + b$



# Computational Units

- The weighted sum of inputs computes a **linear function** of  $x$
- As we already saw, neural networks learn **nonlinear functions**
- These nonlinear functions are commonly referred to as **activations**
- The output of a computation unit is thus the **activation value** for the unit,  $y$ 
  - $y = f(z) = f(w \cdot x + b)$

# There are many different activation functions!

exponential linear unit (elu)

softmax

scaled exponential linear unit (selu)

softplus

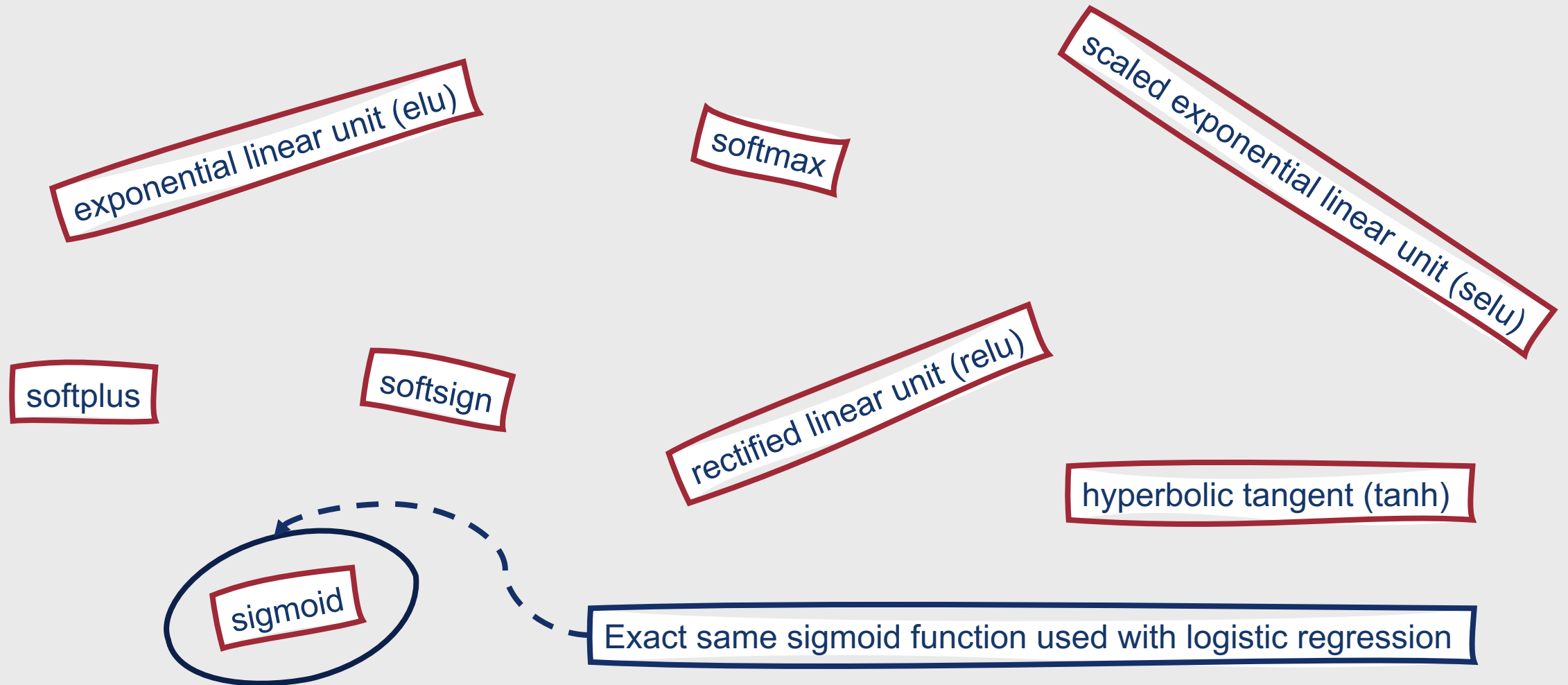
softsign

rectified linear unit (relu)

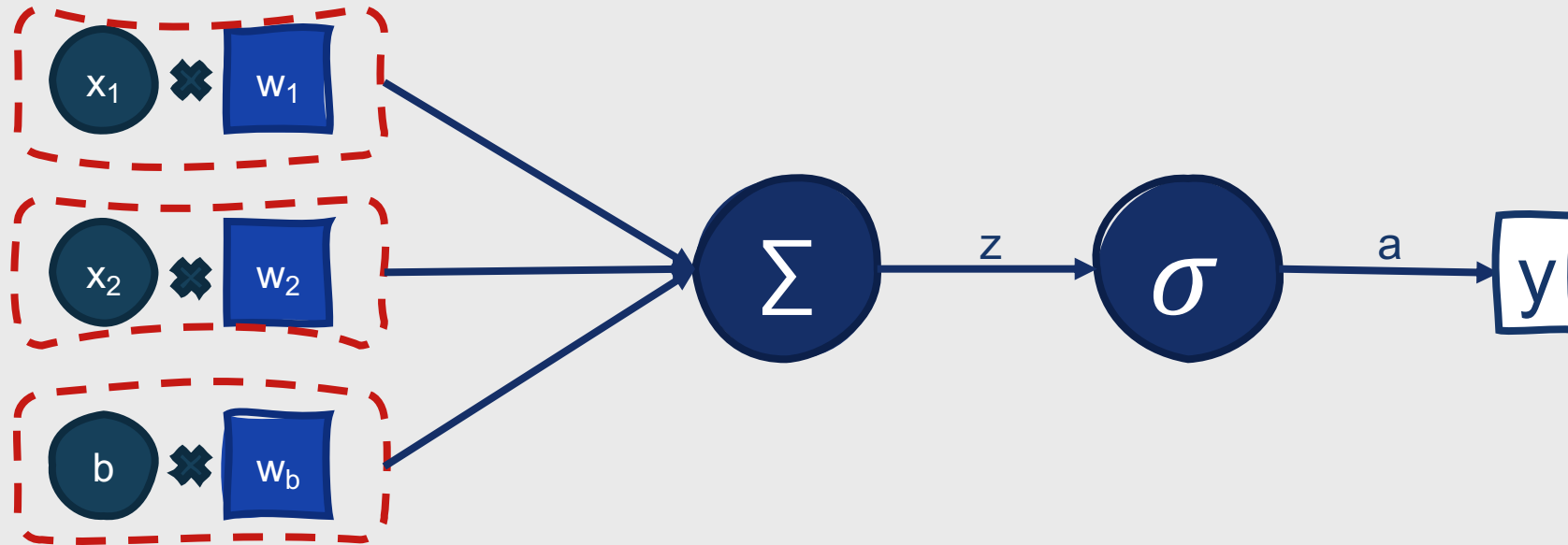
hyperbolic tangent (tanh)

sigmoid

# There are many different activation functions!

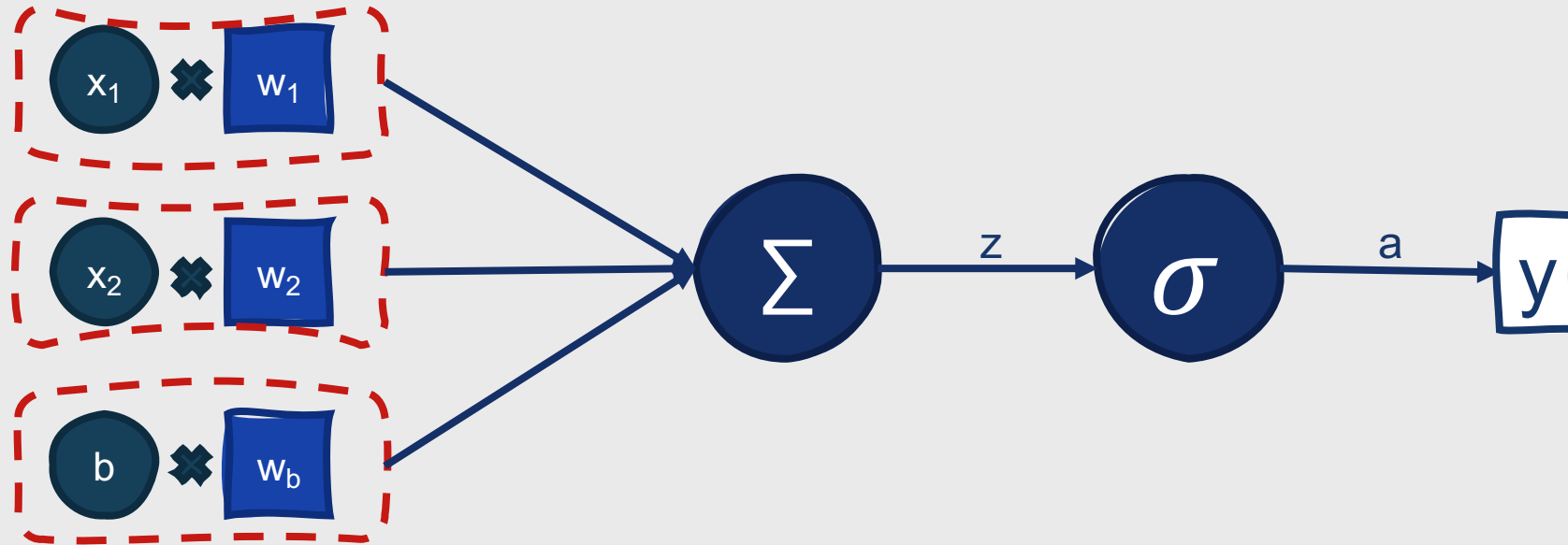


# Computational Unit with Sigmoid Activation





# Example: Computational Unit with Sigmoid Activation

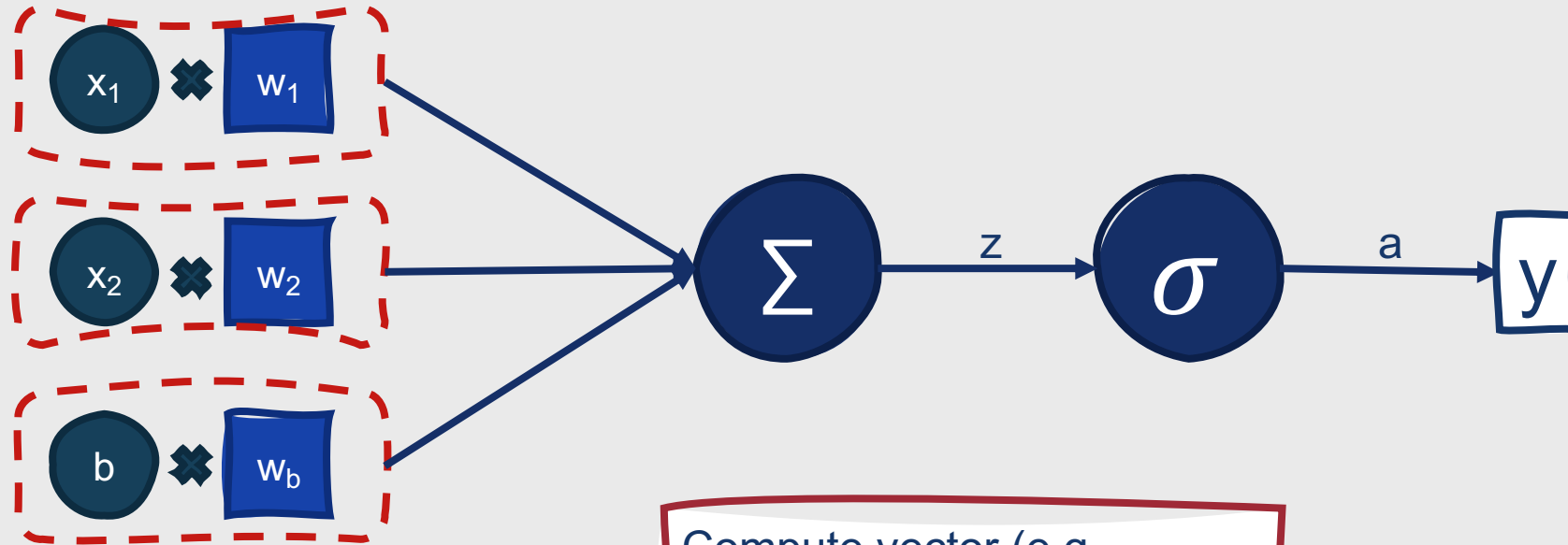


Input: “beautiful brutalist architecture”

Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

# Example: Computational Unit with Sigmoid Activation



Input: “beautiful brutalist architecture”

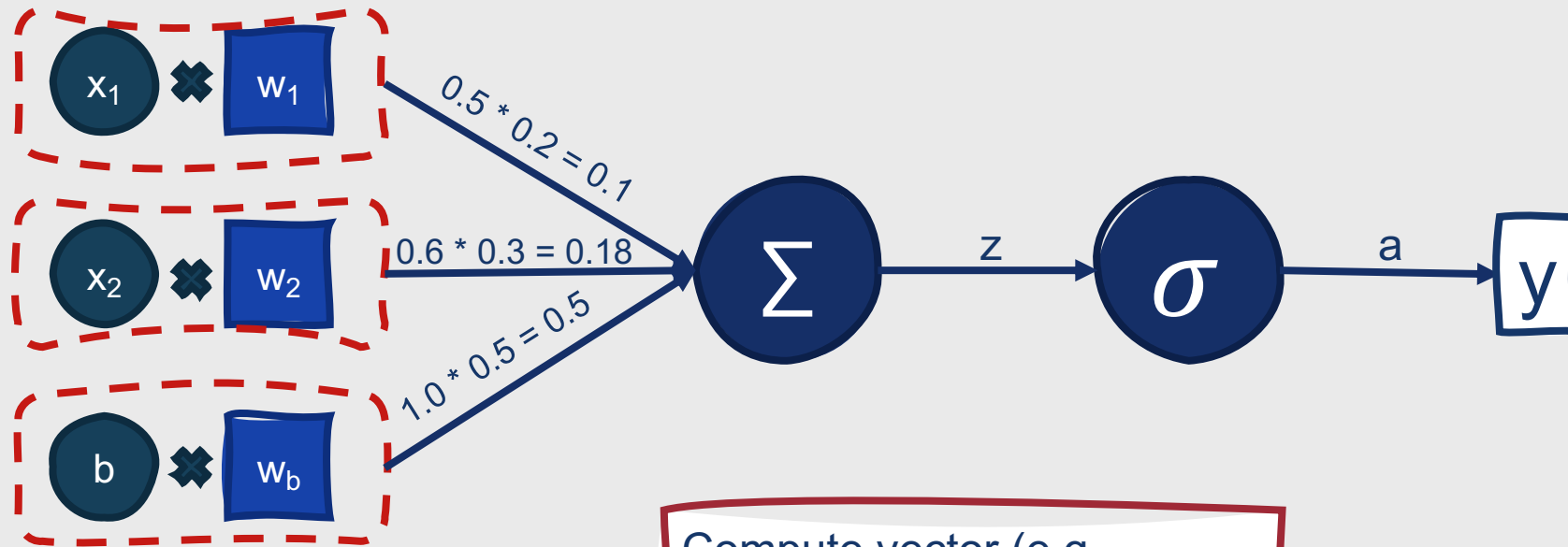
Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for “beautiful,”  
“brutalist,” and “architecture”)

[0.5, 0.6]

# Example: Computational Unit with Sigmoid Activation



Input: “beautiful brutalist architecture”

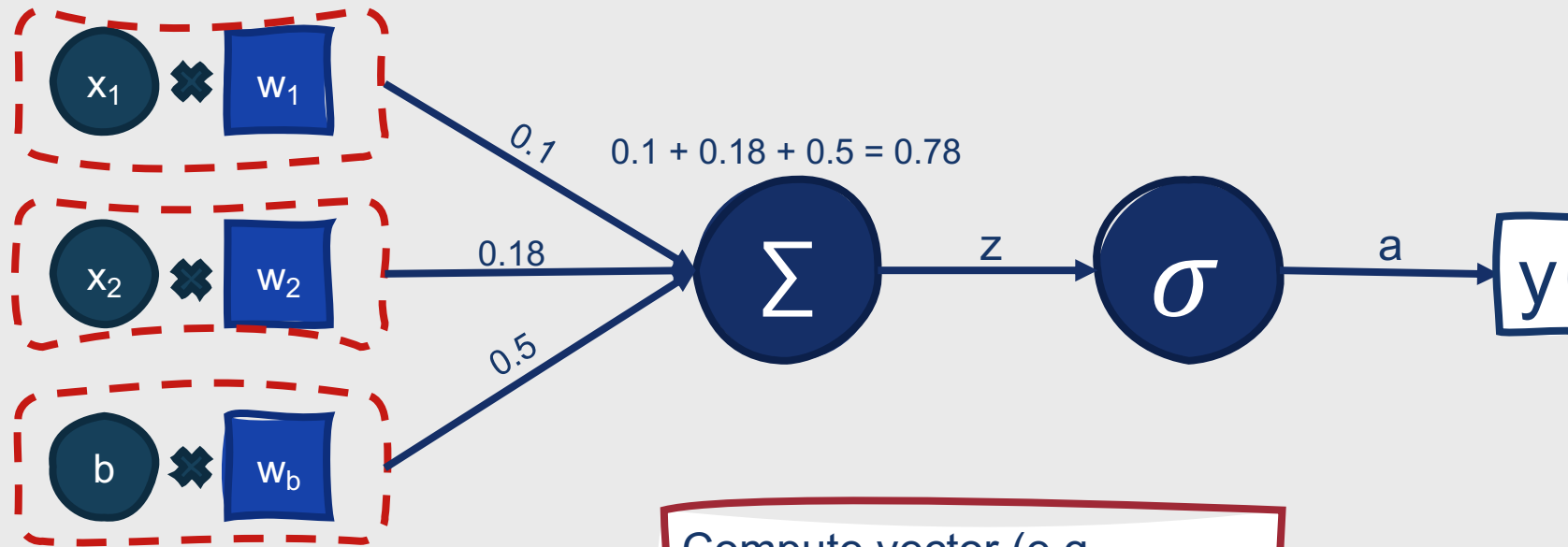
Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for “beautiful,”  
“brutalist,” and “architecture”)

[0.5, 0.6]

# Example: Computational Unit with Sigmoid Activation



Input: "beautiful brutalist architecture"

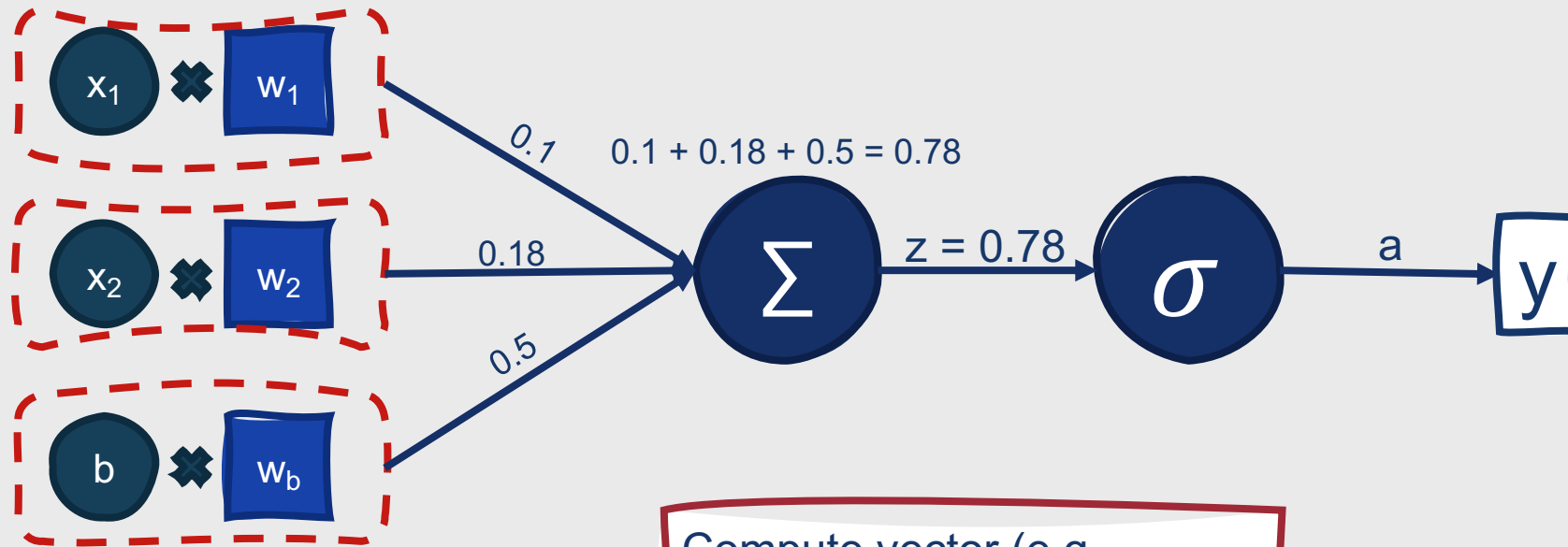
Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for "beautiful,"  
"brutalist," and "architecture")

[0.5, 0.6]

# Example: Computational Unit with Sigmoid Activation



Input: "beautiful brutalist architecture"

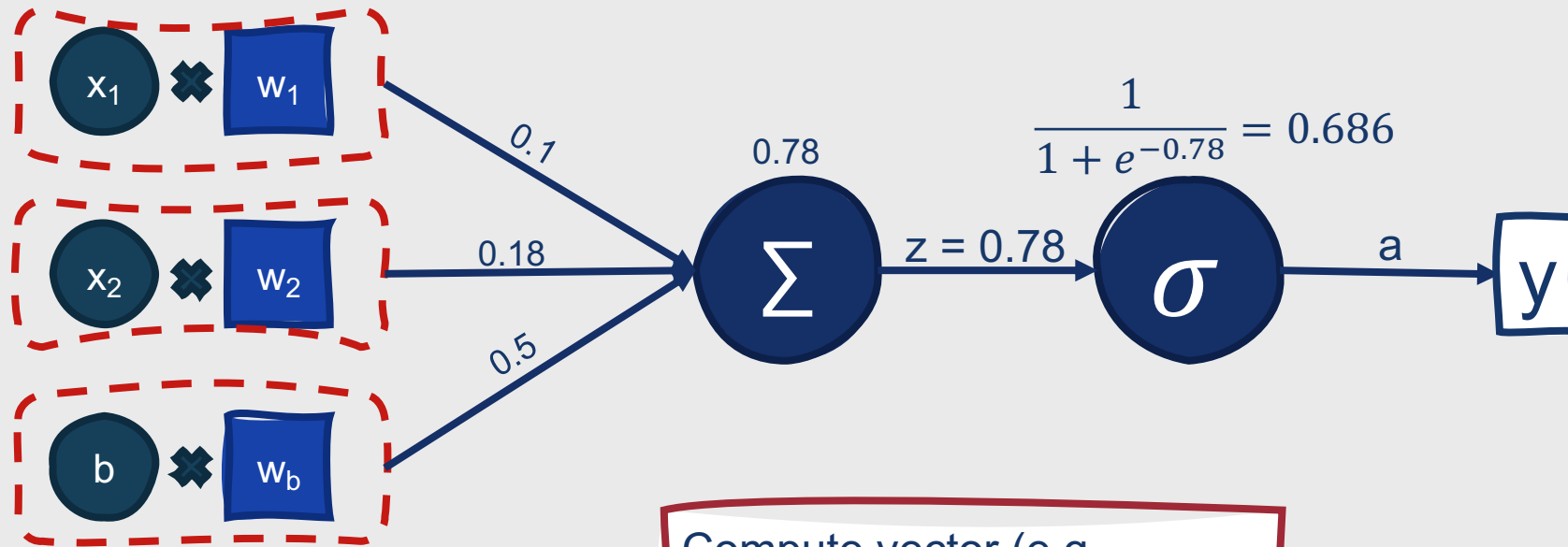
Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for "beautiful,"  
"brutalist," and "architecture")

[0.5, 0.6]

# Example: Computational Unit with Sigmoid Activation



Input: "beautiful brutalist architecture"

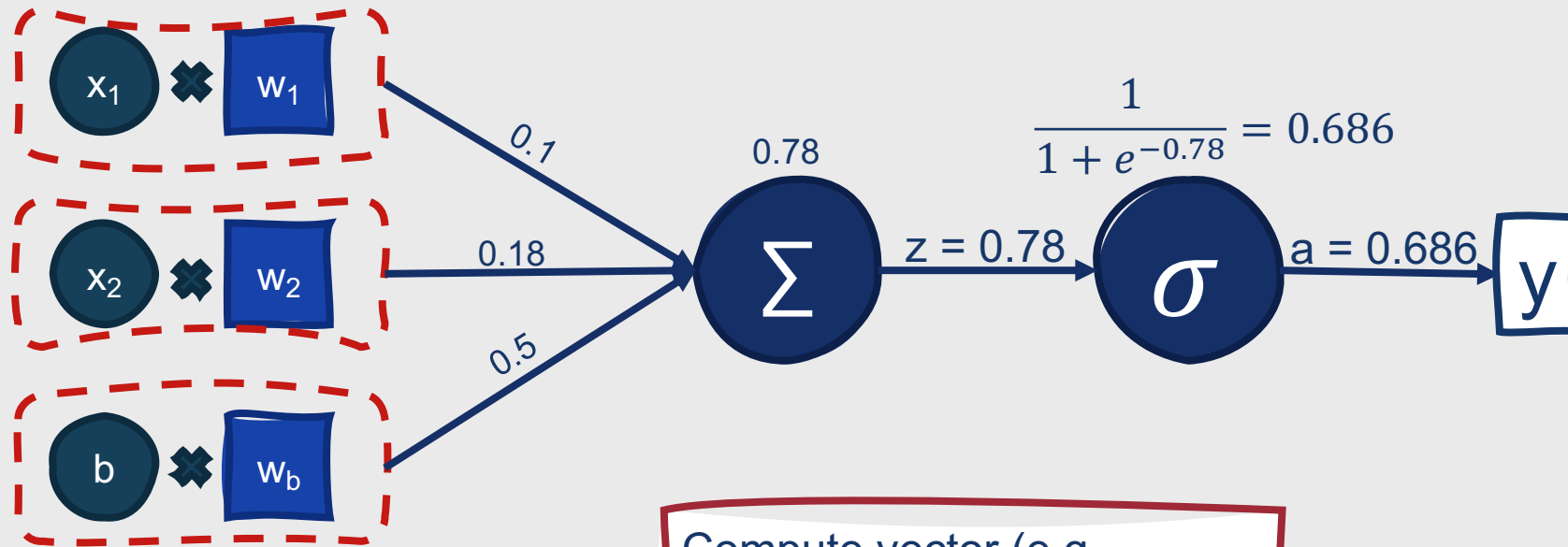
Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for "beautiful,"  
"brutalist," and "architecture")

[0.5, 0.6]

# Example: Computational Unit with Sigmoid Activation



Input: "beautiful brutalist architecture"

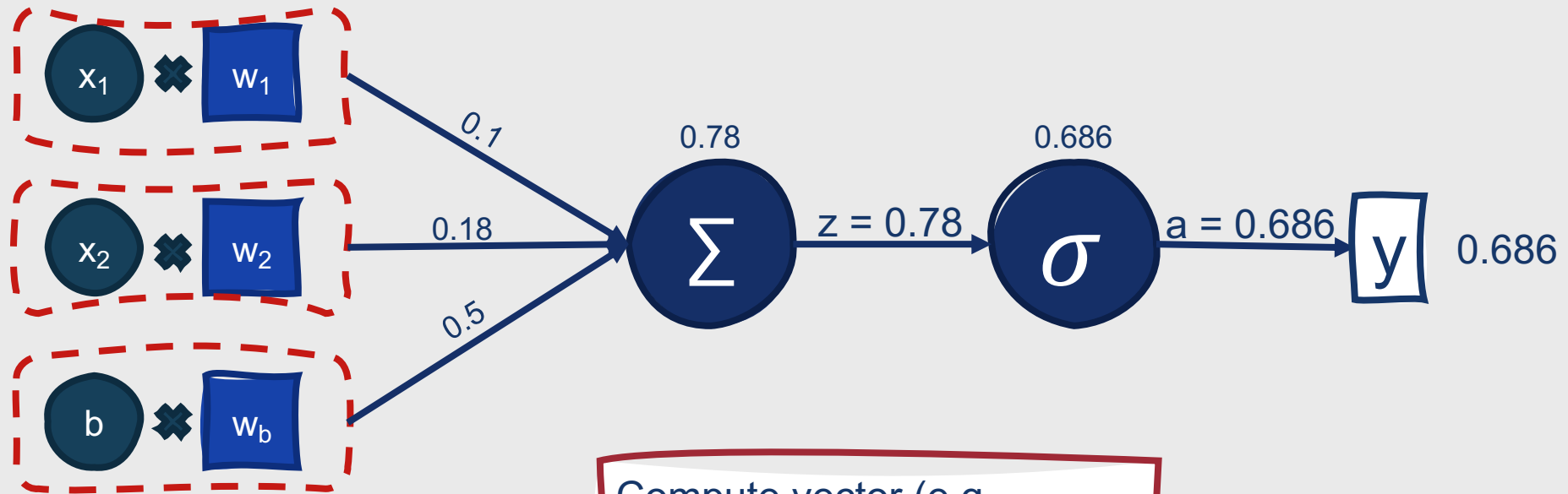
Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for "beautiful,"  
"brutalist," and "architecture")

[0.5, 0.6]

# Example: Computational Unit with Sigmoid Activation



Input: "beautiful brutalist architecture"

Weights (Input): [0.2, 0.3]  
Weight (Bias): [0.5]

Bias: 1.0

Compute vector (e.g.,  
averaged Word2Vec  
embeddings for "beautiful,"  
"brutalist," and "architecture")

[0.5, 0.6]