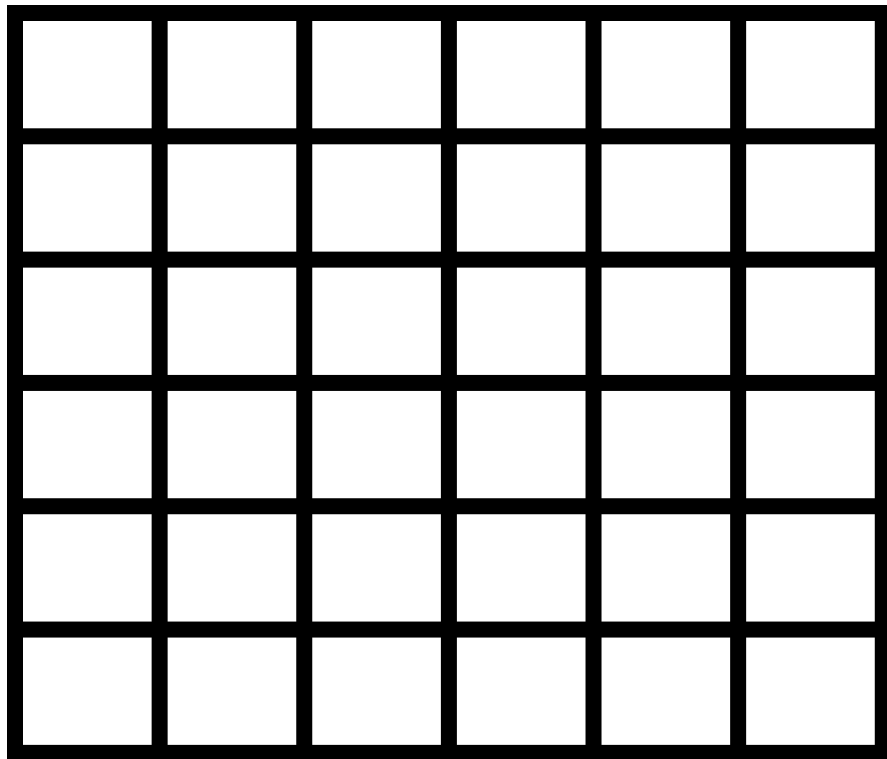
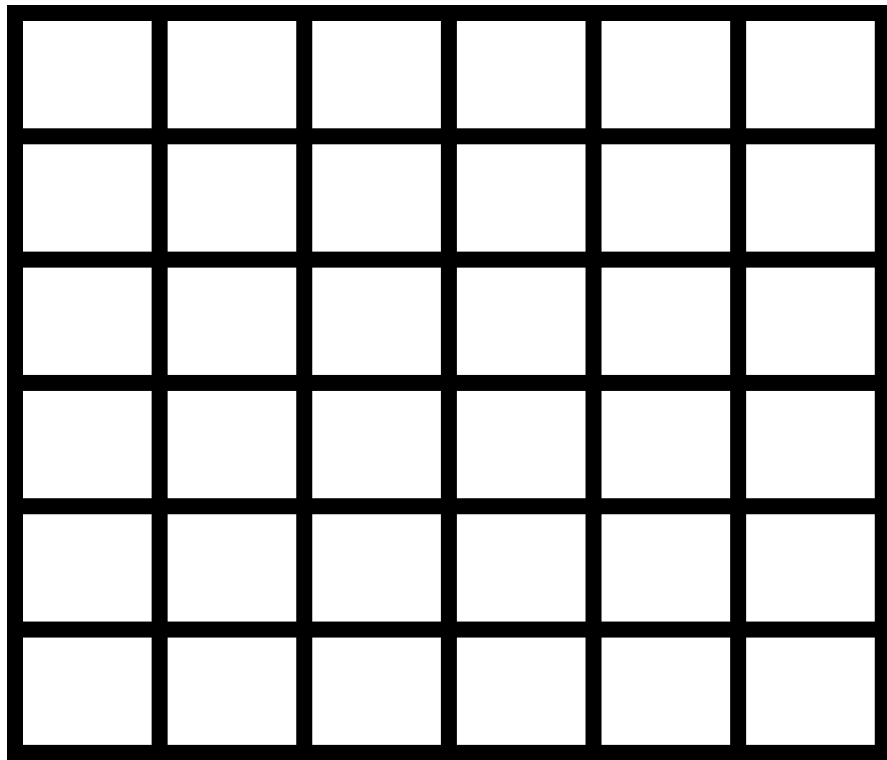


# Convolutional NN for Vision Problems

Image: 2d Matrix of Pixels



## Image: 2d Matrix of Pixels



Old ideas:

- Look within image for interesting areas
- Generate local features from groups of adjacent pixels

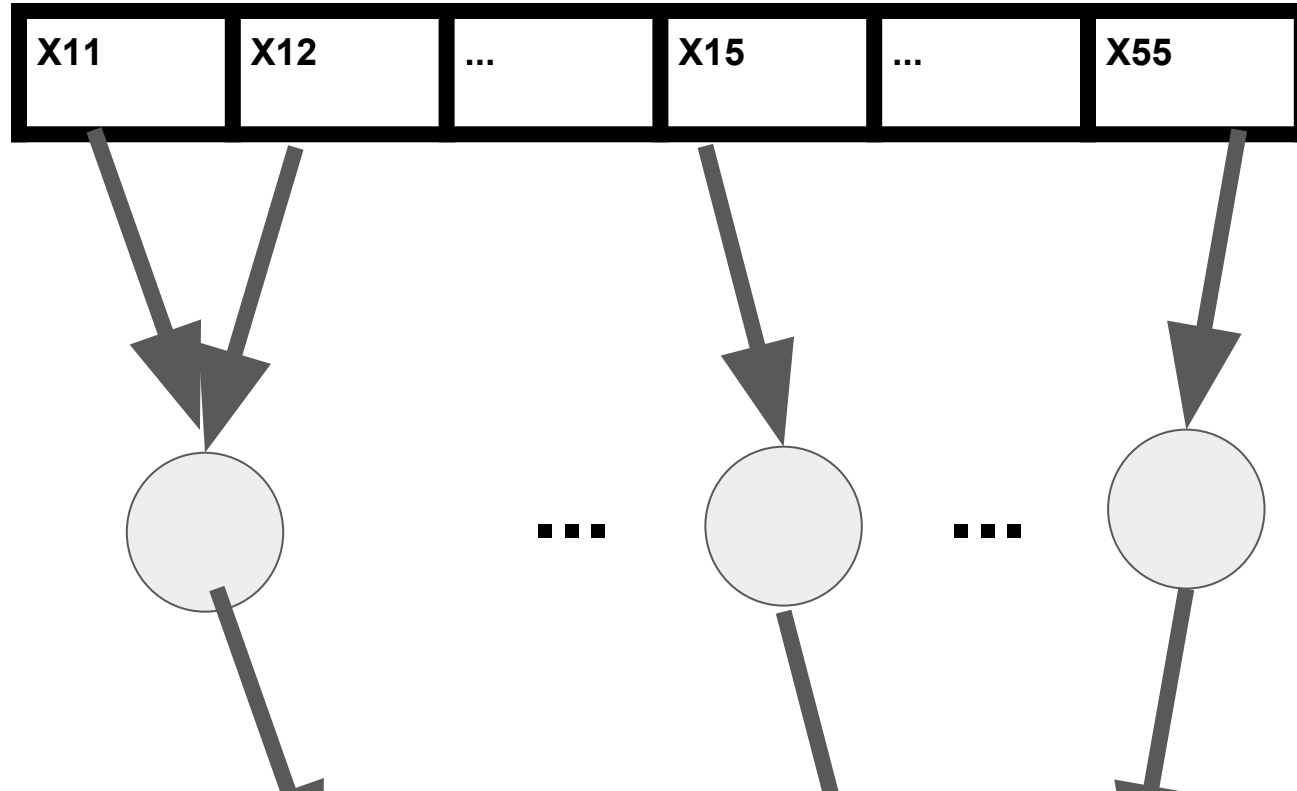
# Image: 2d Matrix of Pixels

X11	X12	...	...	...	X15
X21	X22	...	...	...	X25
...	...	...	...	...	...
...	...	...	...	...	...
...	...	...	...	...	...
X51	...	...	...	...	X55

Neural Nets:

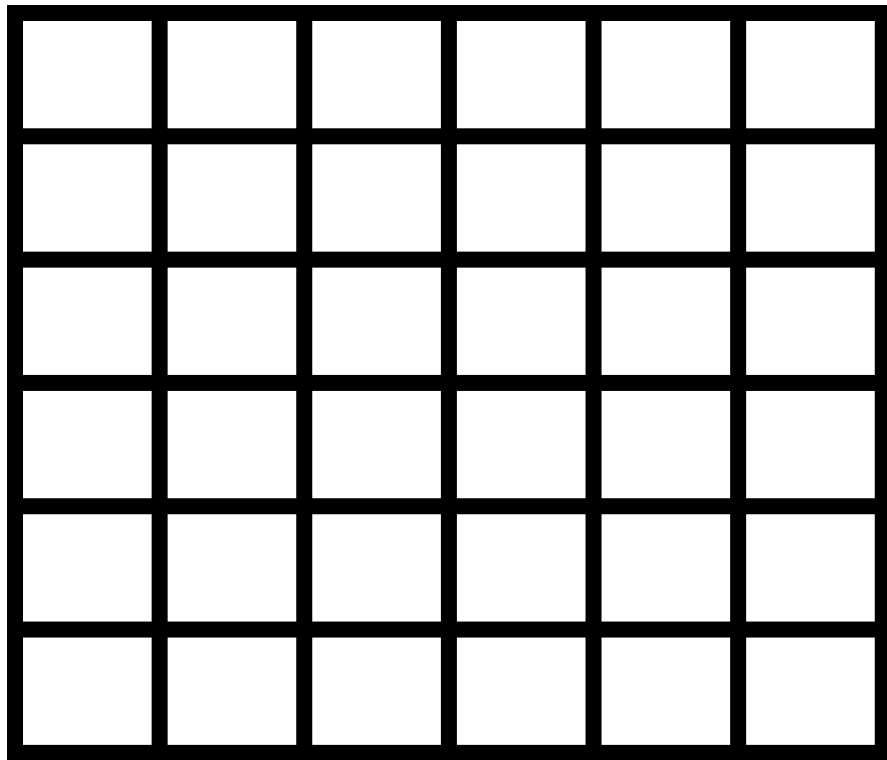
- Pixels are the basic features
- Passed through layers

# Image: 2d Matrix of Pixels; Neural Nets



**How do  
we set up  
these  
layers?**

## Image: NN Layers



### **Idea 1:**

Have adjacent pixels map to a hidden node

Image: 2d Matrix of Pixels

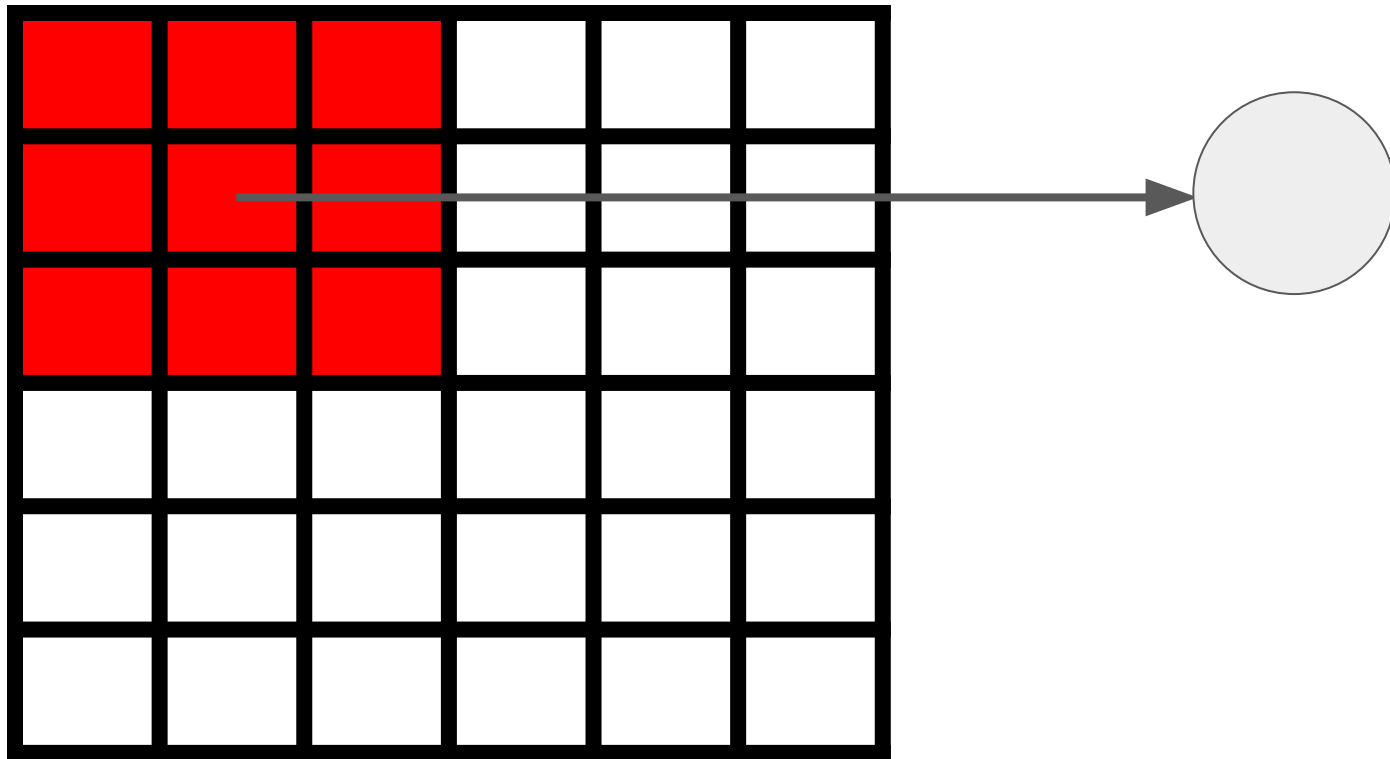


Image: 2d Matrix of Pixels

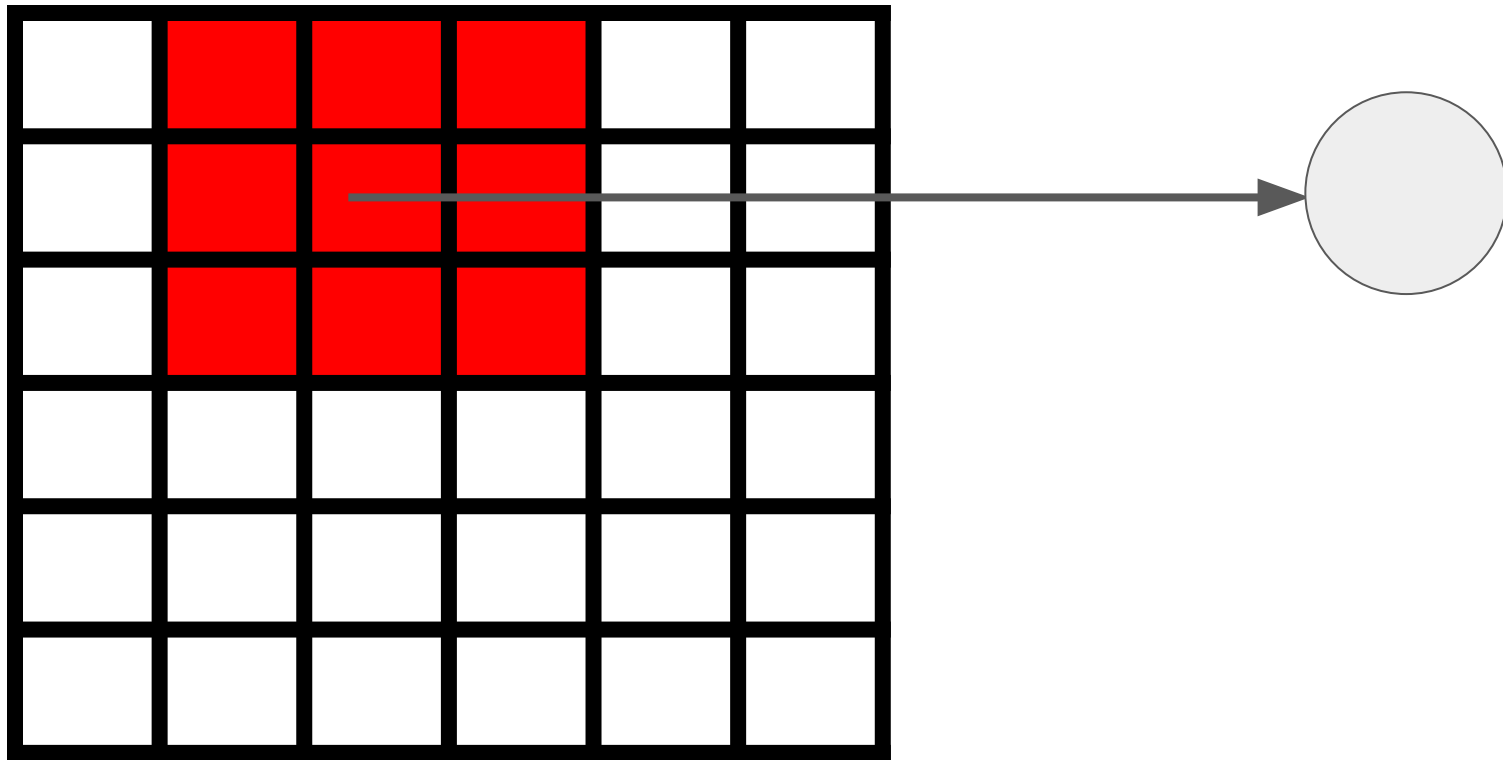




Image: 2d Matrix of Pixels

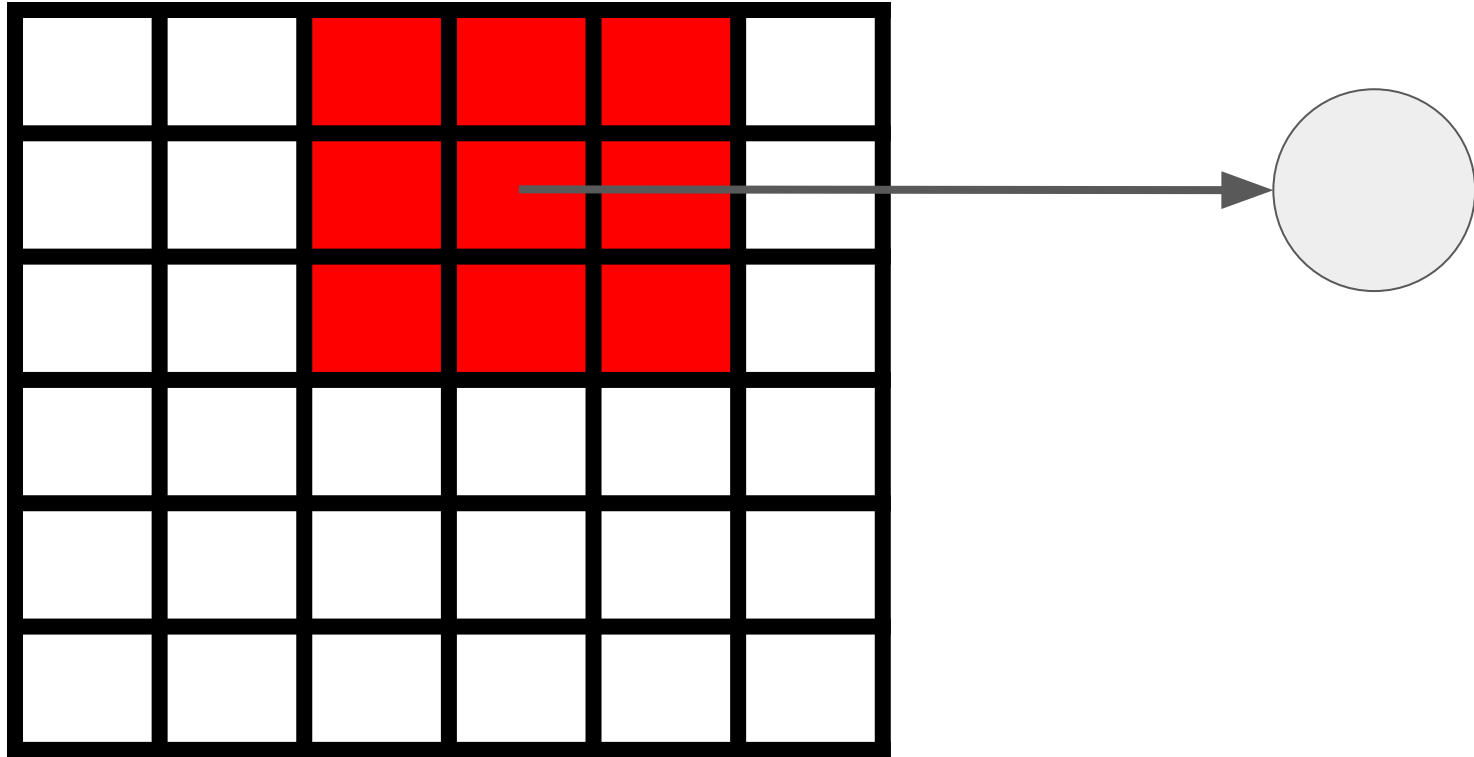


Image: 2d Matrix of Pixels

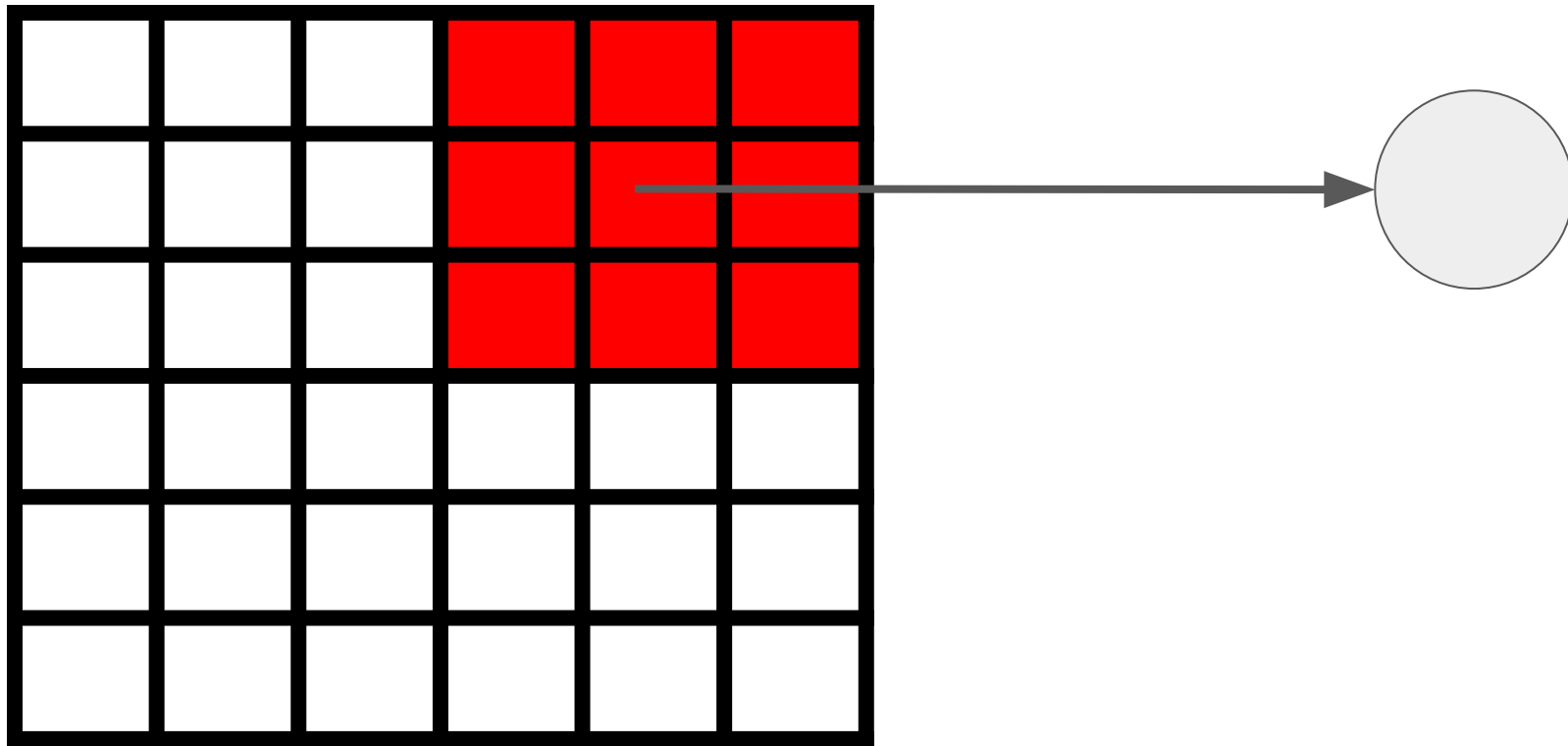
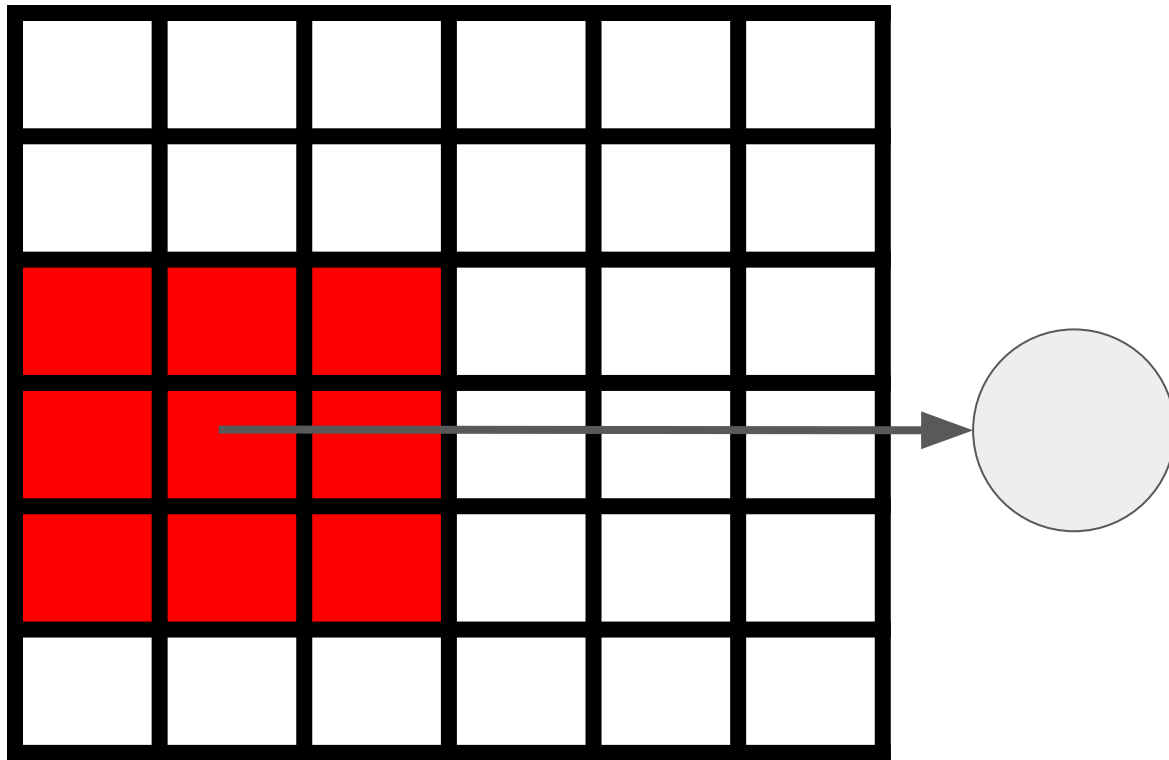
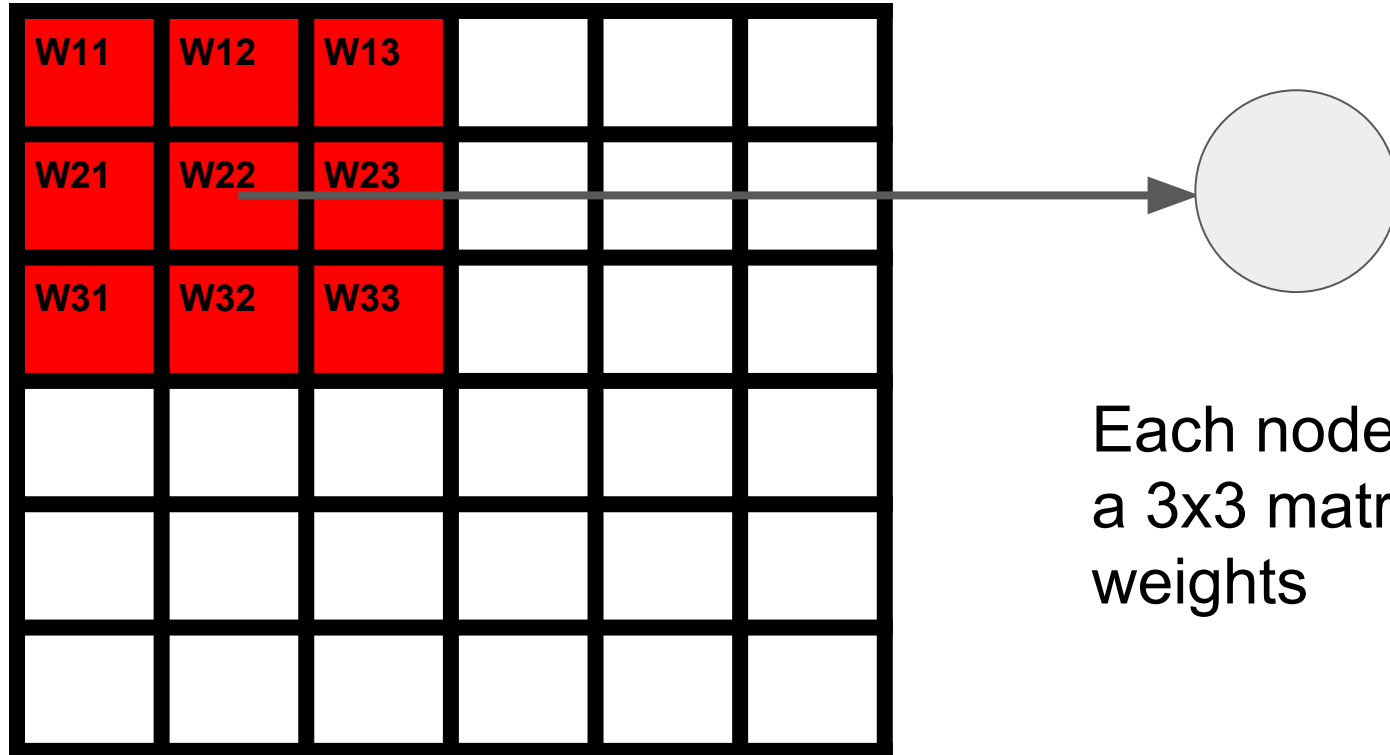


Image: 2d Matrix of Pixels



# Image: 2d Matrix of Pixels, weights



Each node needs  
a 3x3 matrix of  
weights

Image: 2d Matrix of Pixels, *shared weights*

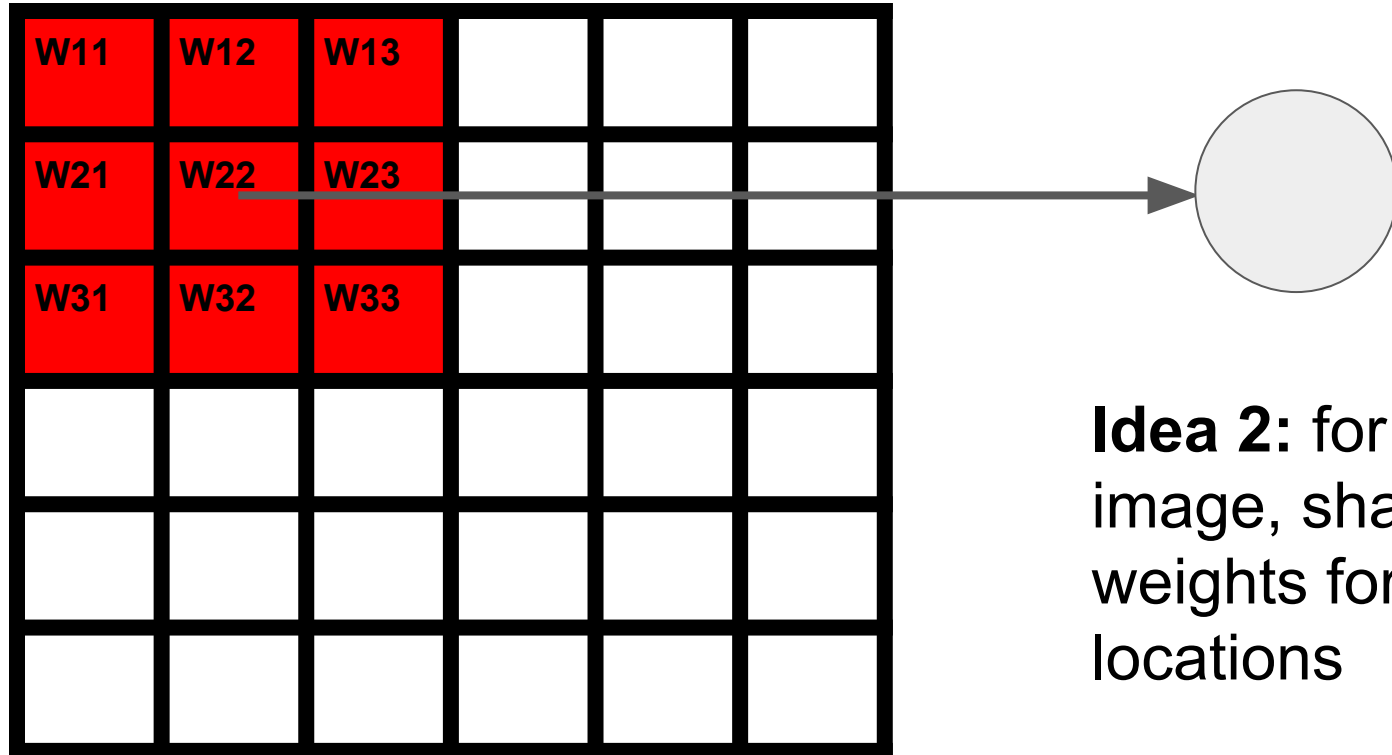
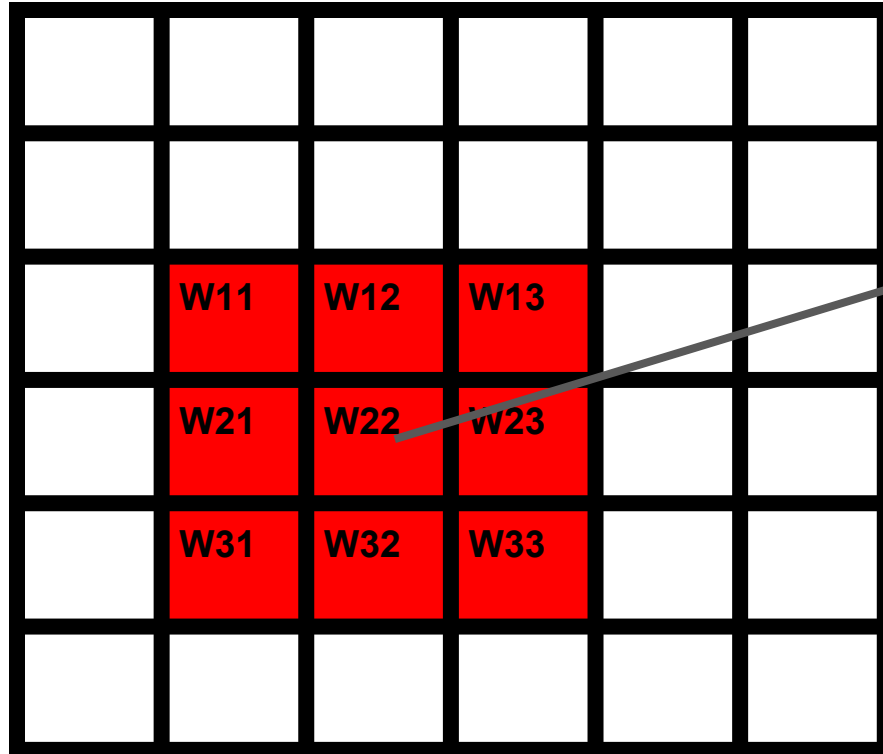


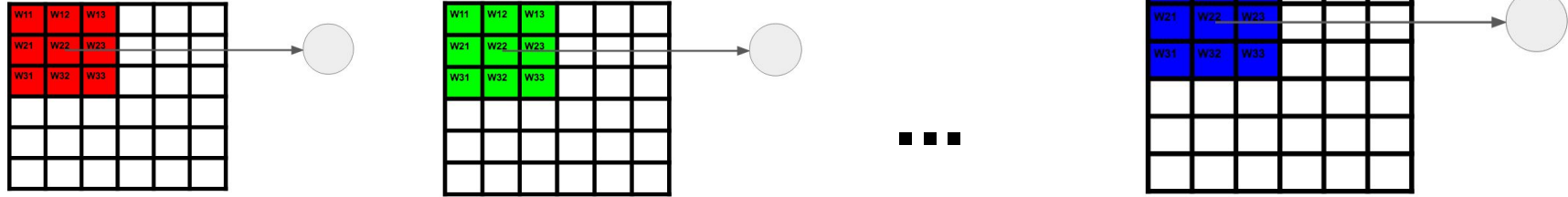
Image: 2d Matrix of Pixels, *shared weights*



**Idea 2:** for entire image, share the weights

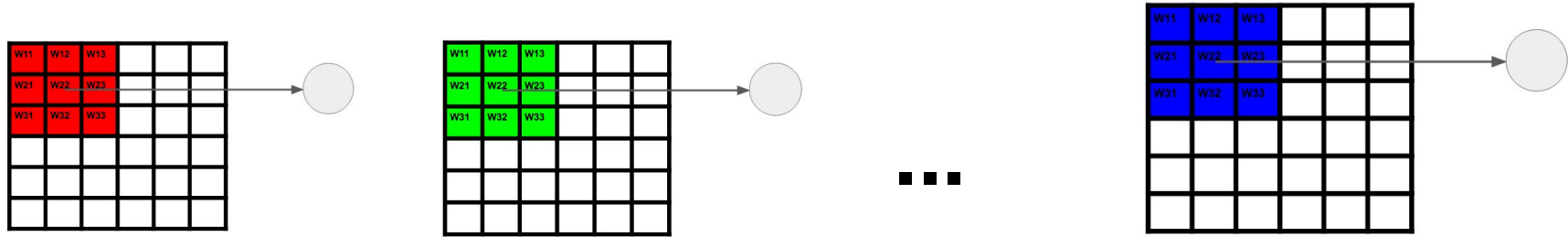
(Call this *feature map*)

# Image: 2d Matrix of Pixels, *feature maps*



**Idea 3:** have many feature maps for the image

# Image: 2d Matrix of Pixels, *feature maps*



**Idea 4:** you can repeat this for hidden layers (feature maps of feature maps)

**Why?** Each feature map output can be arranged like pixels in an image; take output from feature map as an image

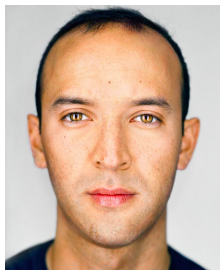


```
theano.tensor.signal.conv2d
```

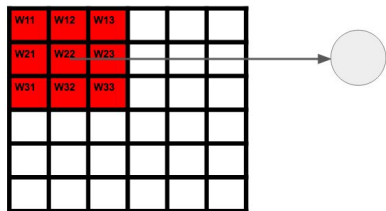
Convenience function for this architecture

Takes as input “4D tensors”, representing the data inputs and weights

# 4D Tensor for data



,



,

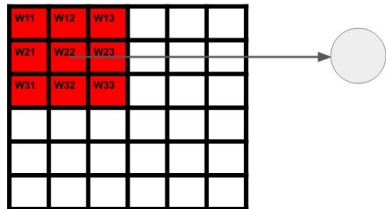
**X**

,

**Y**



,



,

**X**

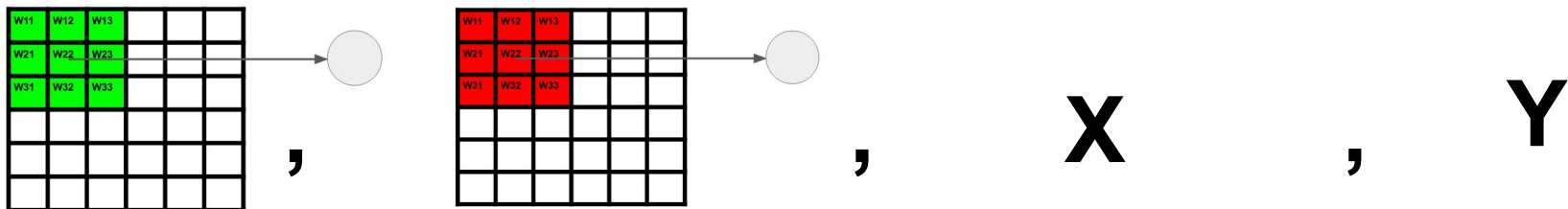
,

**Y**

**4D?**

- (1) Image
- (2) Feature map
- (3) x coordinate of pixel [center]
- (4) y coordinate of pixel

# 4D Tensor for weights



(1) Feature maps out (2) Feature map in (3) x coordinate of pixel [center] (4) y coordinate of pixel

# Convolution layers.

w\_1 =

theano.shared(np.asarray((np.random.randn(\*(featureMapsLayer1, 1, patchWidth, patchHeight))\*0.01)))

w\_2 =

theano.shared(np.asarray((np.random.randn(\*(featureMapsLayer2, featureMapsLayer1, patchWidth, patchHeight))\*0.01)))

w\_3 =

theano.shared(np.asarray((np.random.randn(\*(featureMapsLayer3, featureMapsLayer2, patchWidth, patchHeight))\*0.01)))

...

```
def model(X, w_1, w_2, w_3, w_4, w_5, p_1, p_2):
```

```
    l1 = dropout(max_pool_2d(T.maximum(conv2d(X, w_1,  
border_mode='full'),0.), (2, 2)), p_1)
```

```
    l2 = dropout(max_pool_2d(T.maximum(conv2d(l1, w_2),  
0.), (2, 2)), p_1)
```

```
    l3 =  
dropout(T.flatten(max_pool_2d(T.maximum(conv2d(l2, w_3),  
0.), (2, 2)), outdim=2), p_1) # flatten to switch back to 1d  
layers
```

```
    ...
```