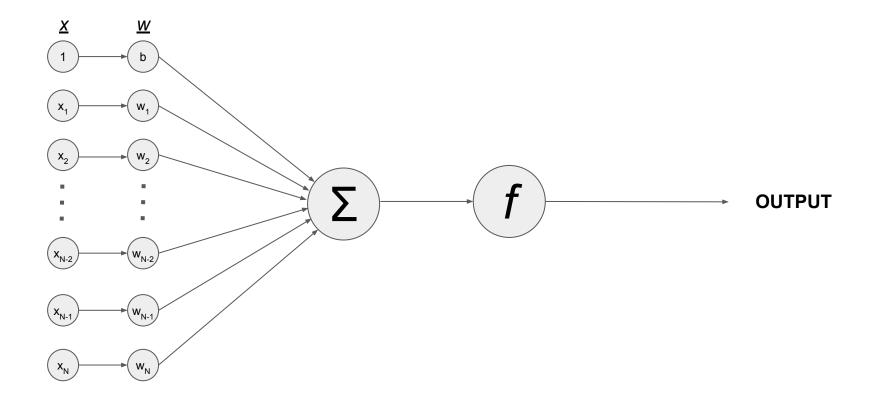
An Application of Machine Learning to Star-Galaxy Classification

Aidan Sedgewick

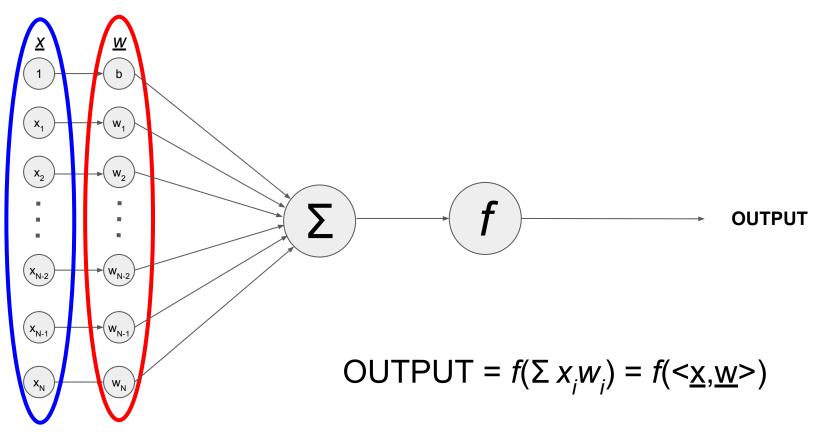
Outline

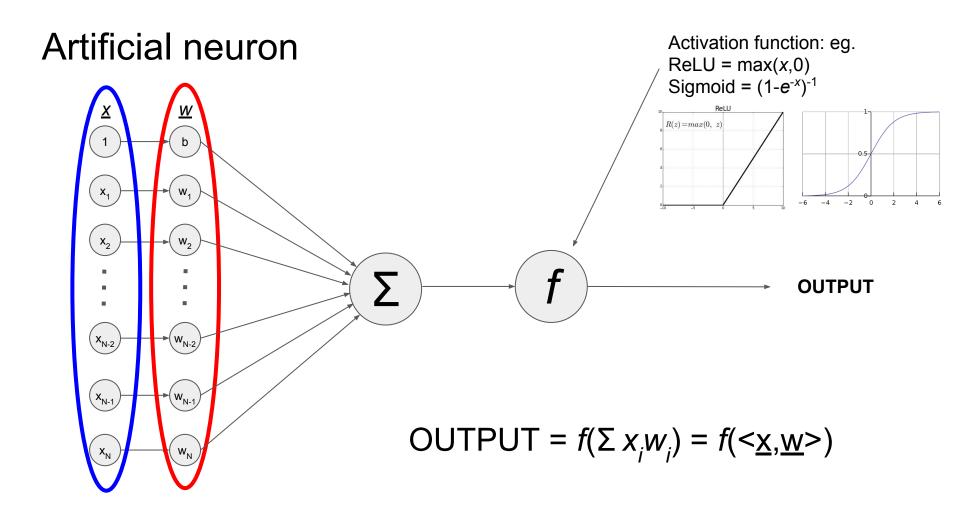
- What is an (artificial) neural network (ANN)?
- ANN vs CNN
- CNN layers: convolution and pooling
- Application to astronomy (PAU survey star/galaxy classification)

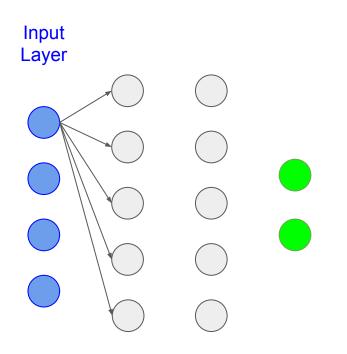
Artificial neuron



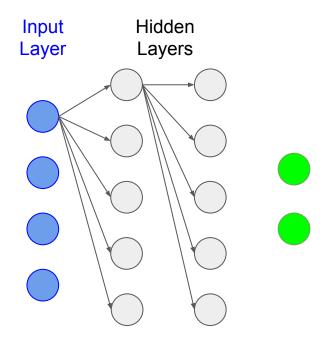
Artificial neuron



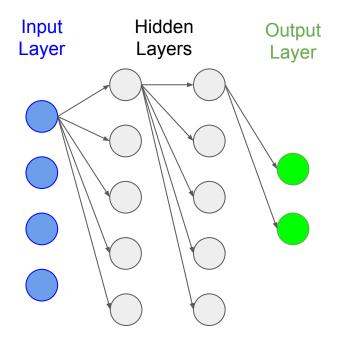




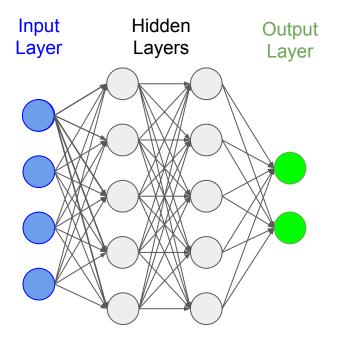
 Can chain artificial neuron outputs to new inputs, and make a 'network'.



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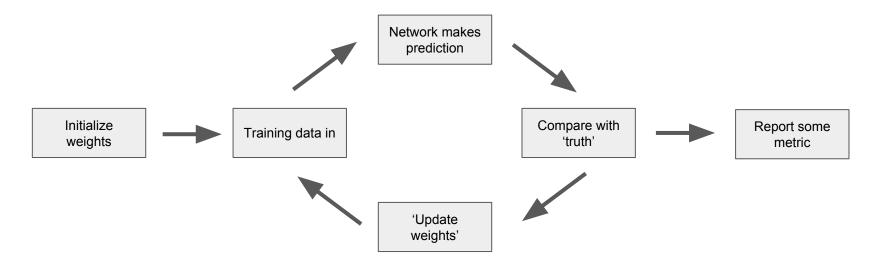


- Fully connected, or 'dense' layers.
- Each neuron has its own set of weights <u>w</u>.

- Number of neurons in a layer?
- Number of layers?
 - \rightarrow Hyperparameters

Training parameters

• How to get a network which outputs the 'correct' answers?



• How many 'epochs' (iterations)?

Convolutional neural network (CNN)

- Often have image data that we'd like to apply ANNs to.
- Need many layers for complex patterns, images with many pixels...
- Huge number of training parameters for ANN!

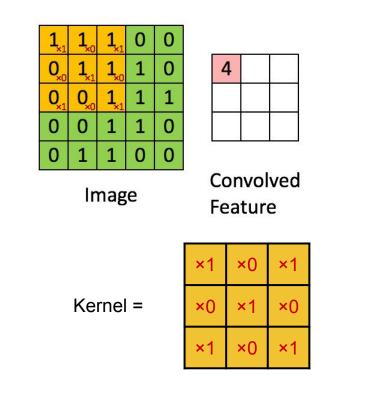
Use convolutional network!

• Made of two main types of layer: convolutional, and pooling layers...

Convolution layers

- Convolutional layers are *not* fully connected.
- Operate *locally*, so can still detect features.

- Kernel: 'Moving window filter'
- Far fewer parameters to be learned.
- No activation function here.



Pooling layers

10	8	3	7
4	3	8	1
2	11	4	3
9	5	5	1

8

5

10

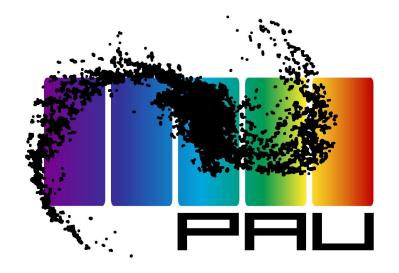
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- "Serve to reduce dimensionality" coarse graining of images.
- Max pooling or average pooling are commonly used.

- Usually no trainable parameters in these layers.
- Usually apply activation function *f* at this stage.

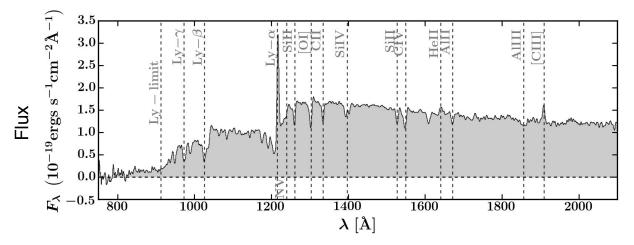
PAU Survey

- Physics of the Accelerating Universe Survey
- Provide accurate "photometric redshift" measurements for up to 30 million objects.



Redshift Estimates: spectro-z

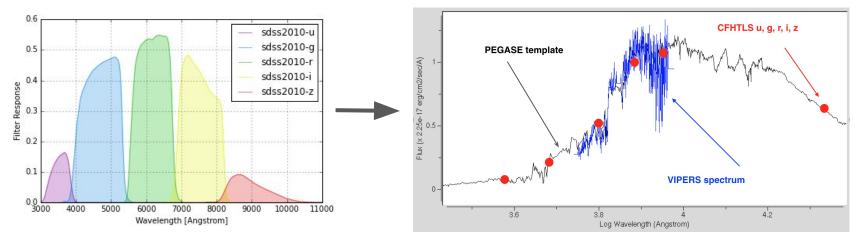
- Most accurate way to obtain redshift estimates is with high-resolution spectra ('spectro-z')
- *Extremely* time consuming, but error of 0.01%



CANDELS-VIMOS

Redshift Estimates: photo-z

• Can obtain rough redshift estimates (~5% error) with few images in different bands ('photo-z') very quickly.

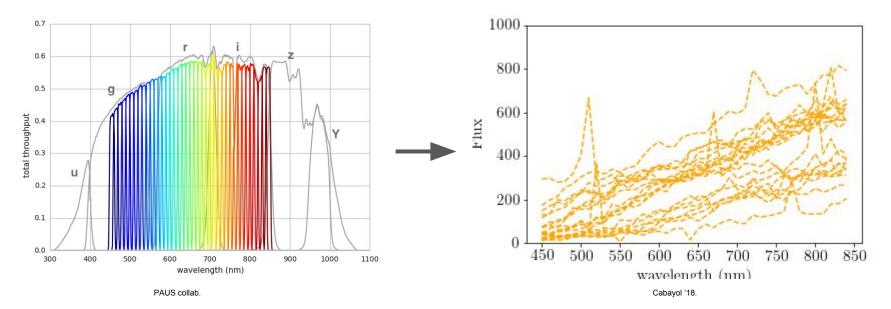


github.com/dkirkby/speclite

Giorgio Manzoni

PAUS

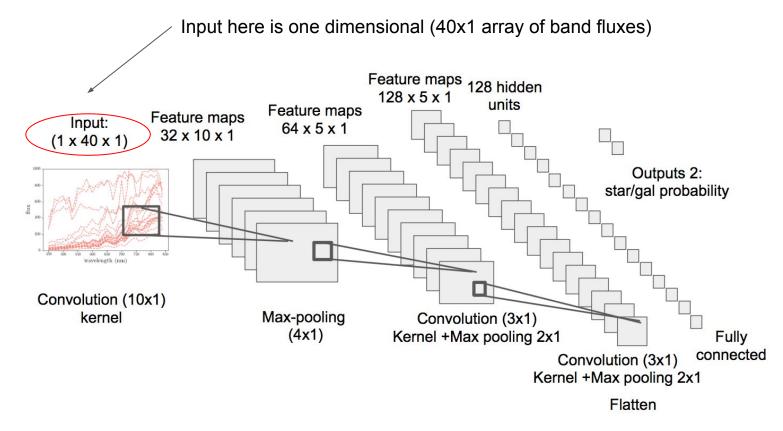
- Traditionally have ~6 filters
- PAU Cam has 40 narrow band filters!
- Effectively low resolution spectra.



Applying NNs to PAUS

- First, need to determine what type of object, to use relevant template.
- Can we also use the band fluxes to determine this?
- Problem in this paper is star-galaxy classification.
- Use previously imaged fields (*Hubble* data) as a training set 1.2M objects.
 1.1M galaxies, 30k stars → *strong* class imbalance

Network design



Performance metric

- How to evaluate the performance of the classifier?
- Defining a galaxy detection as a 'positive' result:

	Classified as galaxy	Classified as star
True galaxy	TP	FN
True star	FP	TN

$$TPR = \frac{TP}{TP + FN}$$

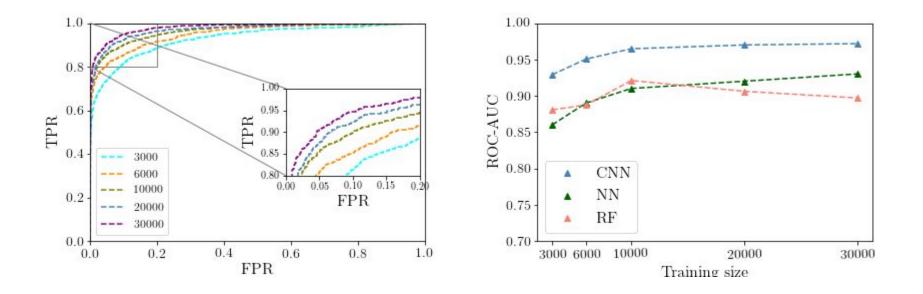
'Fraction of true galaxies correctly classified as galaxies'.

$$FPR = \frac{FP}{FP + TN}$$

'Fraction of true stars falsely classified as galaxies'.

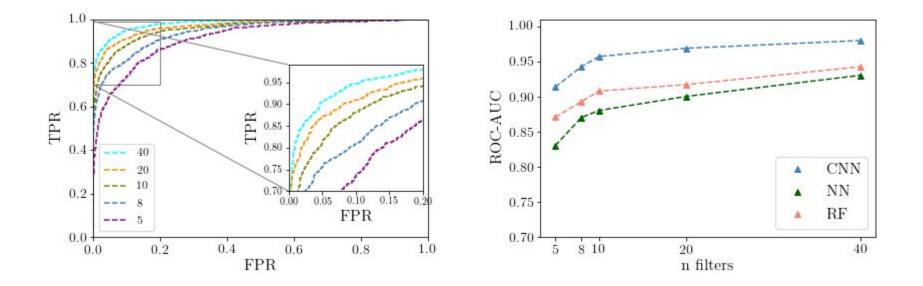
ROC Curves - training size

- For some acceptable *FPR*, what is the obtained *TPR*?
- Area under curve is also a useful measure of performance.

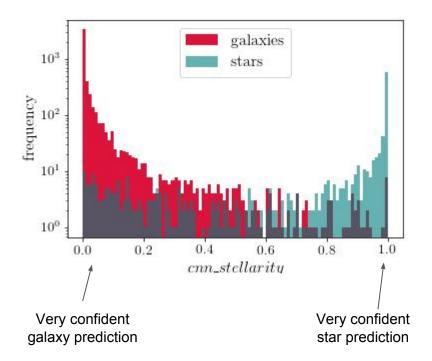


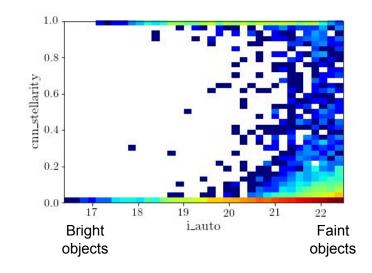
ROC Curves - input size

• Combine consecutive filters to simulate broader bands.



Output distribution





Result

• 99% purity, for 98% completeness.

"Of all the galaxies in the sample, we detect 98% of them. 99% of what we label galaxies truly are galaxies".