

That is  
definitely a picture  
of a **Penguin!**

What ??!!!



# Troubles with Deep Learning

Paul Gader



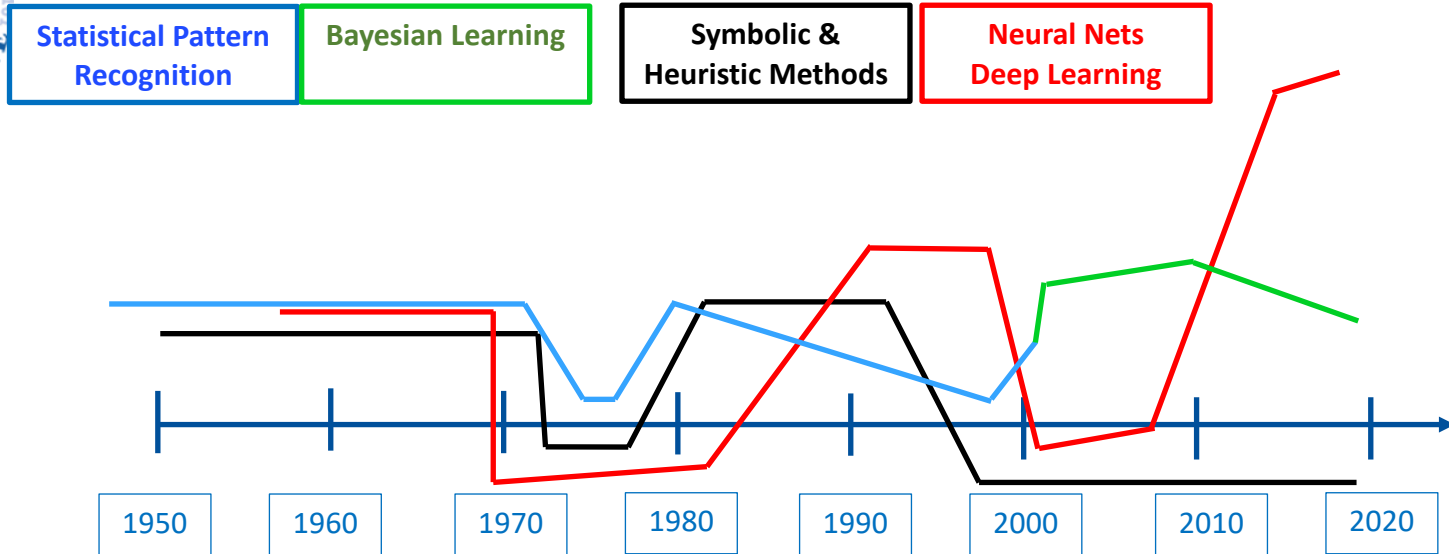
## Artificial Intelligence

**Computer Systems that demonstrate intelligent behavior or perform tasks requiring perception**

- |   |  |
|---|--|
| <ul style="list-style-type: none"><li>• Medical Diagnosis,</li><li>• Speech &amp; Handwriting Recognition,</li><li>• Natural Language Processing,</li><li>• Human-Computer Interaction,</li><li>• Big Data Sets</li></ul> | <ul style="list-style-type: none"><li>• Computer Vision,</li><li>• Sensor Networks,</li><li>• Signal/Image Analysis</li><li>• Remote Sensing</li><li>• Playing Games,</li><li>• Robotics</li></ul> |
|---|--|



# Simple Illustration of History of AI



**Artificial Neural Networks (ANNs) driving field currently**

**Three waves of ANN investigations**



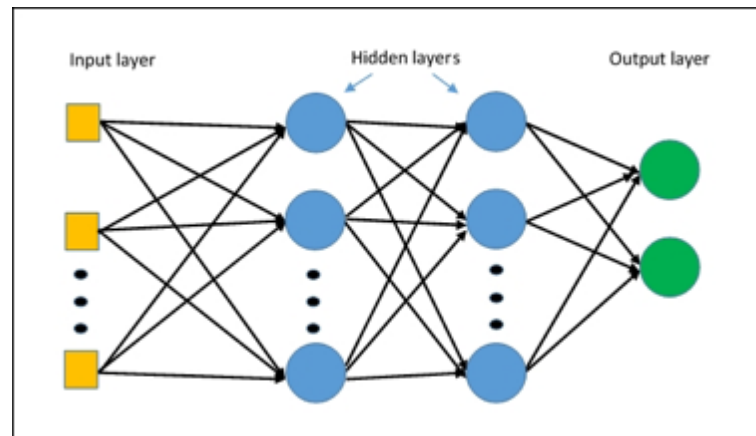
# Everyone "knows" ANNswith Deep Learning is Great



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- Deep Learning = many layers
  - Weights
  - Connections
- Some Astounding Acccomplishments Reported on in Press...





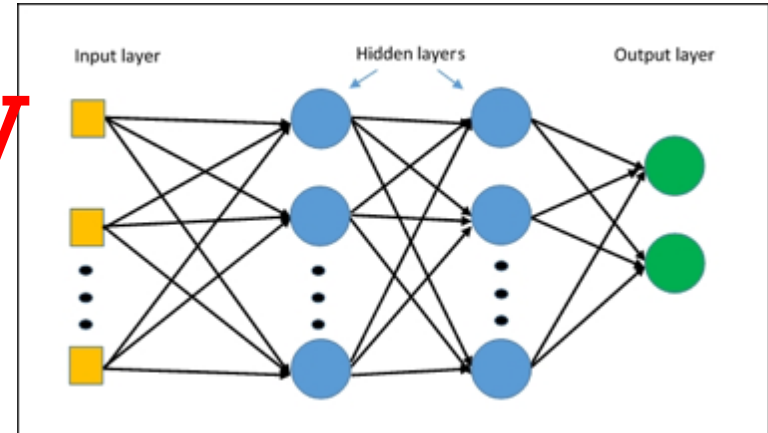
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So  
What Could Possibly  
Go Wrong?





# Three Waves of Neural Network Computation



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- Wave 1
  - 1950s and 1960s
  - Single Layer Networks called Perceptrons
  - Ended When Minsky & Papert showed Single Layer Networks couldn't do XOR
- Wave 2
  - 1980s and 1990s
  - Associative Networks, CNNs, Multi-Layer Perceptrons (2 hidden layers max)
  - Started with Hopfield and Rumelhart and McClellan
  - Faded away and primarily replaced with SVM and Bayesian techniques
- Wave 3 is Now



## Basic Drivers of Three Waves

- High Performance Processors
- New Architectures (e.g. GPUs)
- Better Memory Management (e.g. Caching)
- Large Data Sets (Internet, Sensors, Hard Drives)



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# Problem

- ANNs Learn Complicated Functions in High-Dimensional Space
- Dimension = the Number of Weights in Network
- Often Millions of Dimensions!
- Hard to control



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## Problem (cont.)

- Neural Networks are **not Robust**
  - Large changes to inputs can produce no changes in outputs
  - Small changes to inputs can produce large changes in outputs



## We Want Competency Aware ANNs

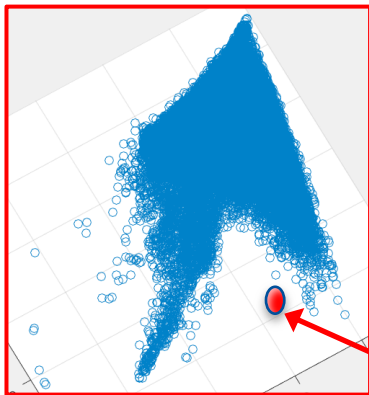
- Competency Aware ANN can identify when
  - Input is **valid or invalid**
- Many ANNs can't identify invalid inputs
  - Why not just look at distance to training set?



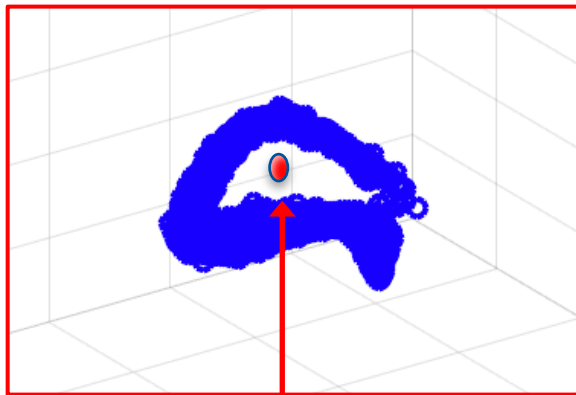
## We Want Competency Aware ANNs

Consider these 3D Projections of High-Dimensional Hyperspectral Data

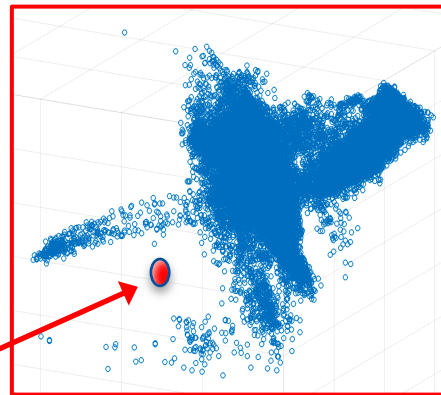
California Shrubs



Coral Reef



Mars



Are these points close?



# Examples



# The Early Days

## Handwritten Address Recognition

~ 1990-1993

2001 Lake Rd	
Youngstown, OH 44174	

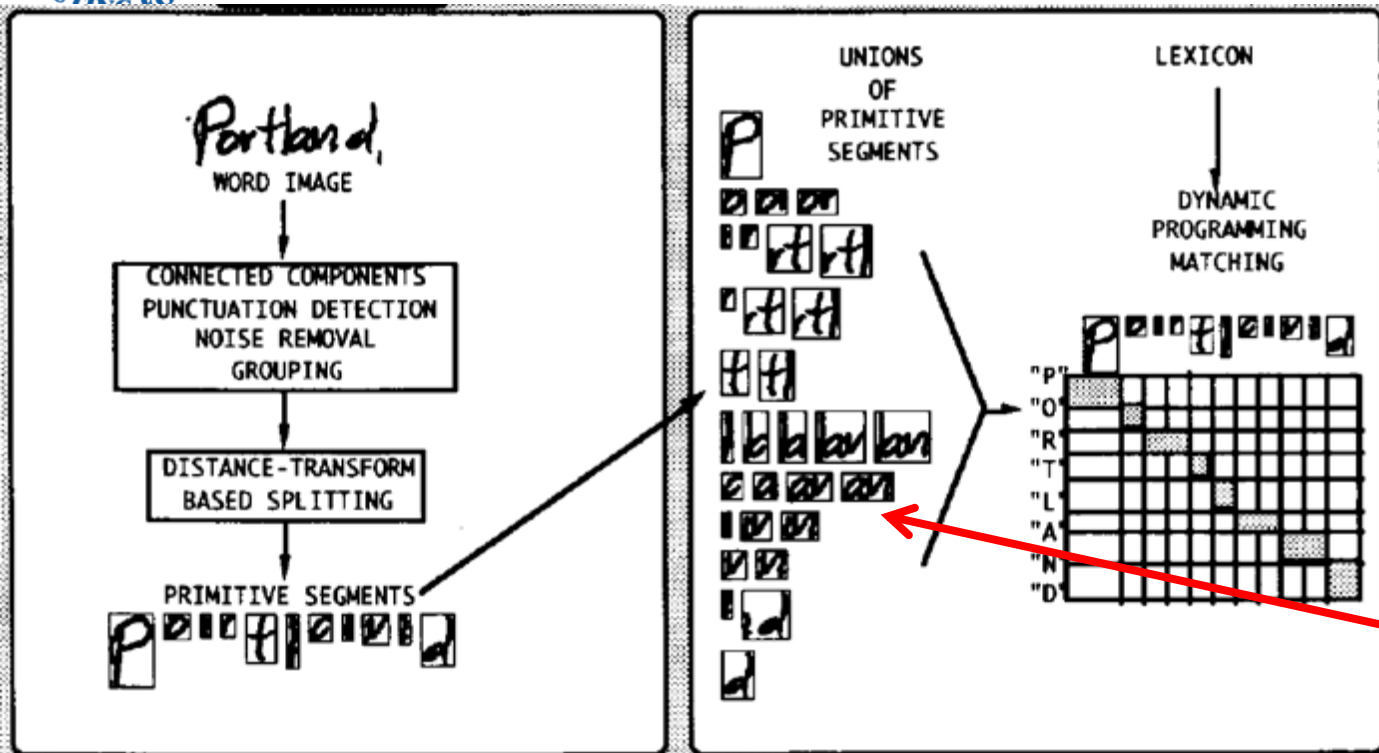
Cowleville

Ha a H 9 U Co





# The Early Days Handwritten Word Recognition ~ 1990-1995



Oversegment & Match  
Word Images to Strings

Find best combination of  
image segments to  
match string

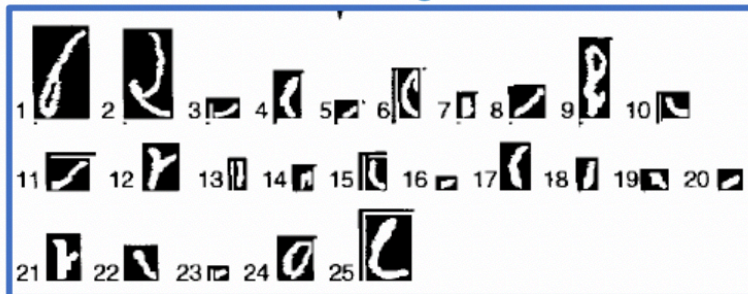
Many  
Non-Characters  
=  
Invalid Inputs



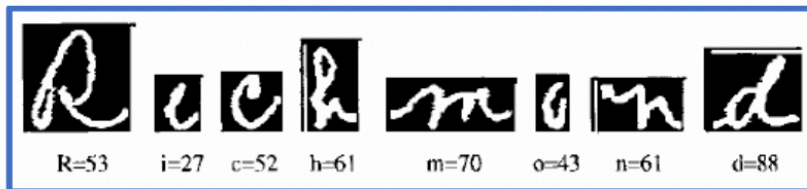
Another  
Look

Richmond

Primitive Segments



Best Unions of Primitives to Match "Richmond"



Best Unions of Primitives to Match "Richmond"





Self-Organizing Map  
(SOM)

# The Early Days

## Handwritten Word Recognition

~ 1990-1993



Do you really look  
like them?

**Classifier**

→ Classification

→ Confidence

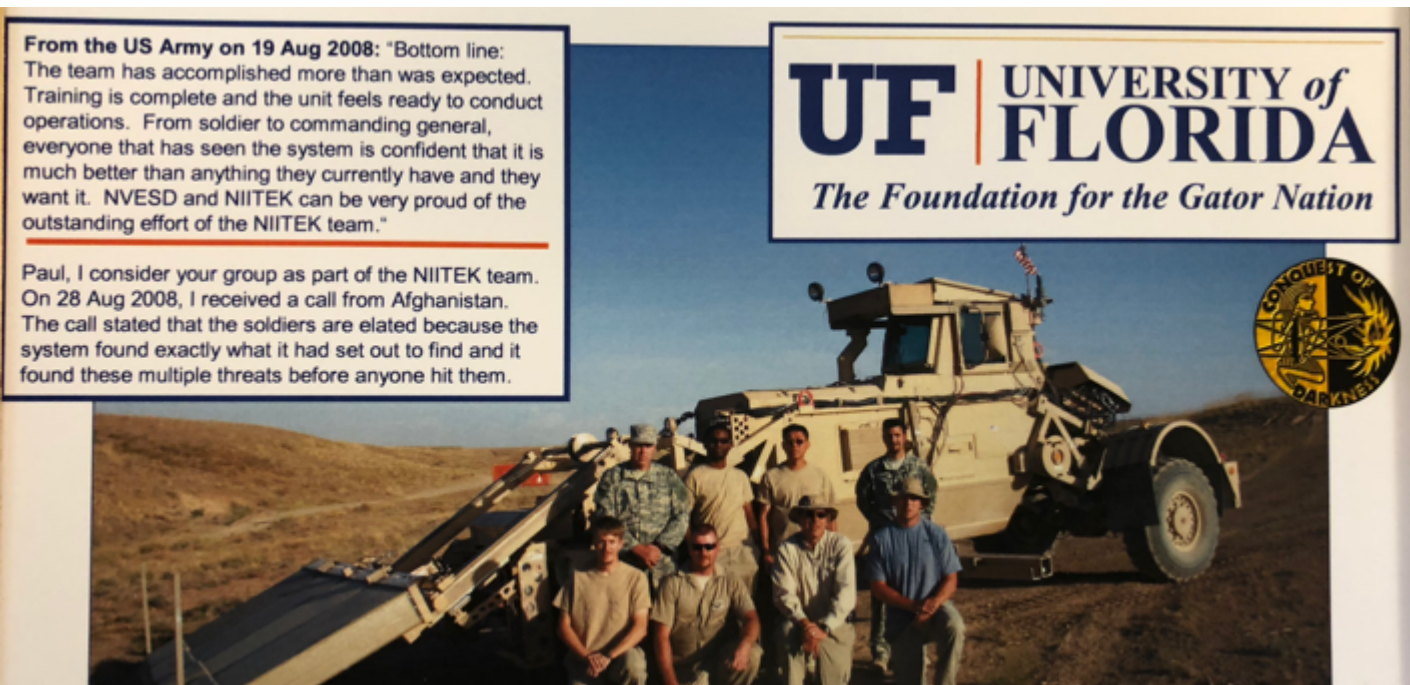
**Better for Word Recognition**  
**Worse for Character Recognition**



# Progression to LM / IED

## Fielded in 2008

### TV: "Bomb Hunters: Afghanistan"



Extensive  
Algorithm Testing

**SOM-based  
Method Best**

and

was first to be  
fielded

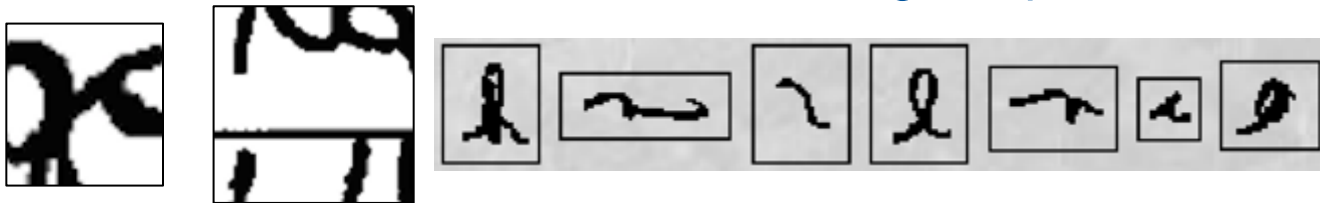
(by my former post-  
doc and colleague  
Hichem Frigui)



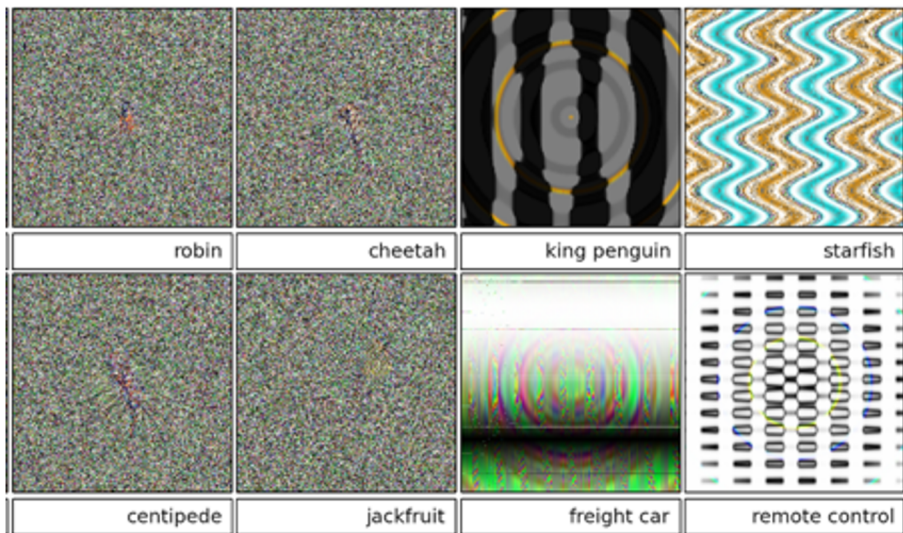
# Invalid Input Examples

Neural Networks can make confident classification decisions on “Garbage” Inputs

■ Natural



■ Concocted





# The Elephant in the Room

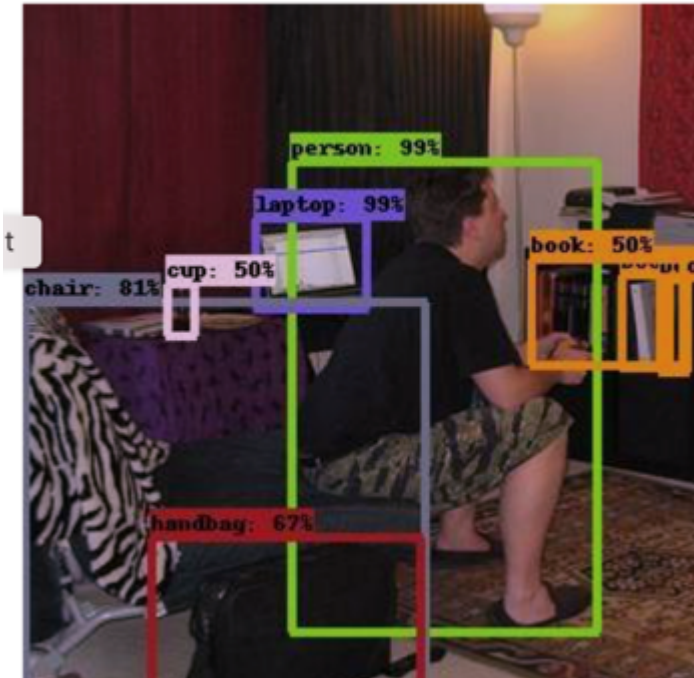
Amir Rosenfeld<sup>1</sup>, Richard Zemel<sup>2</sup>, and John K. Tsotsos<sup>1</sup>

- A state-of-the-art object detector detects multiple images in a room
- An image of an elephant can be put in the room and cause many problems

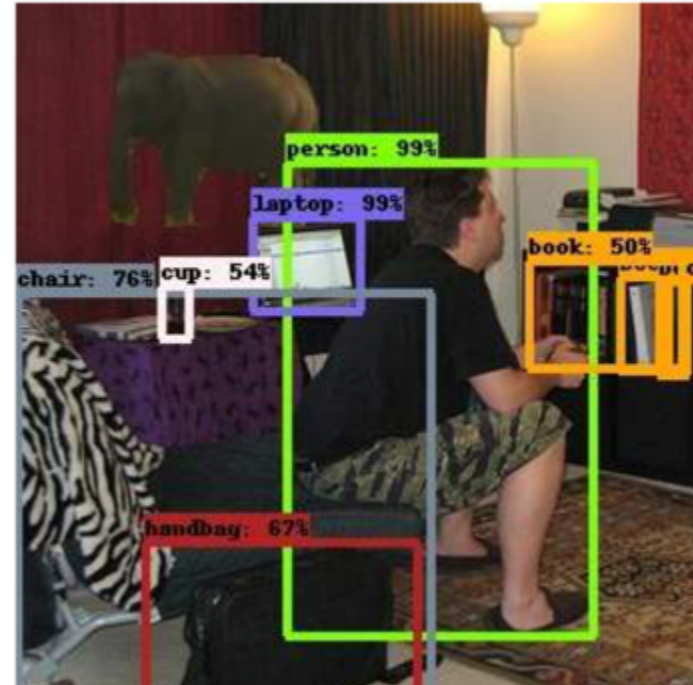


## Elephant in Room Examples

Original Image



Elephant 1 – No Change



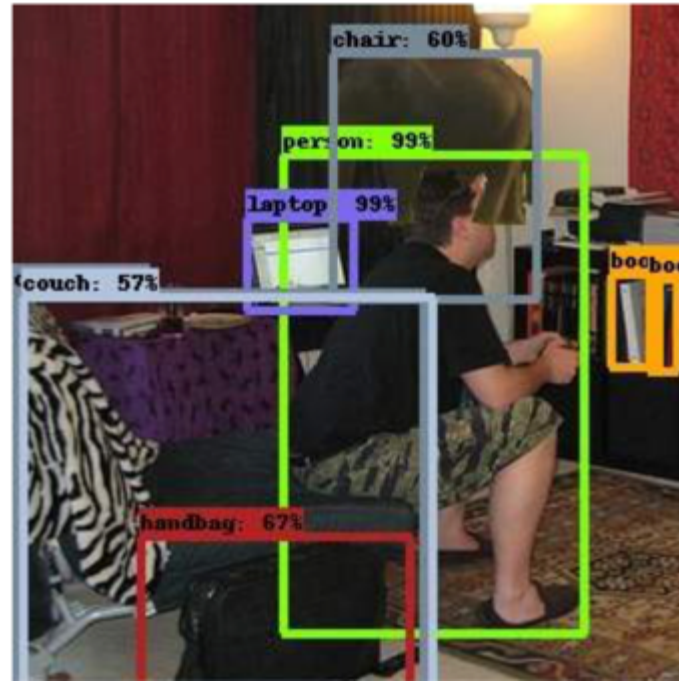
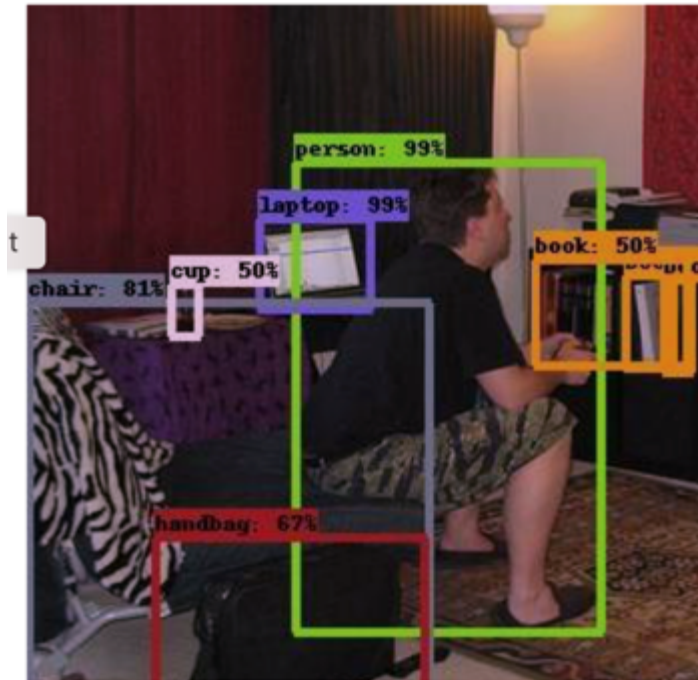




## Elephant in Room Examples

Original Image

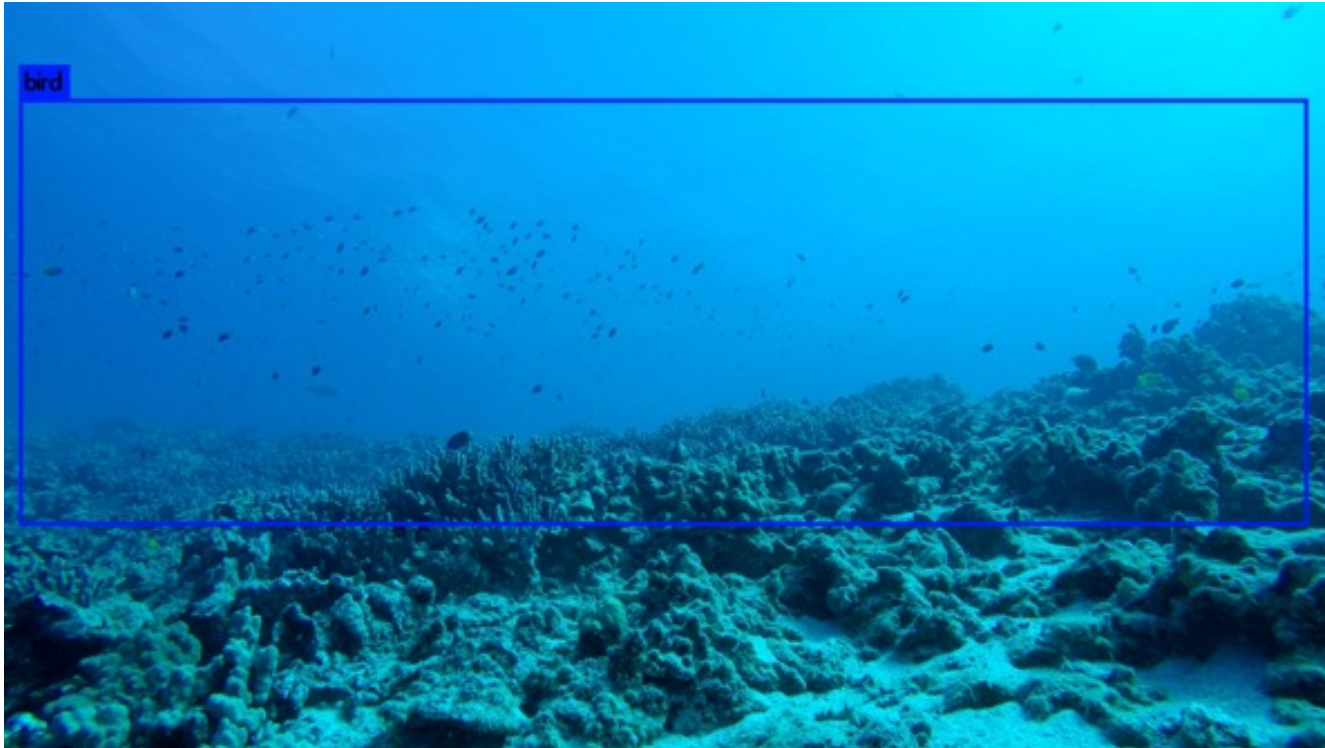
Elephant 3  
Chair becomes Couch  
Cup Disappears





Blue Box Called a Bird

My PhD student Ron Fick ran  
Yolo ANN (Well-known) on Underwater Video





# So, what should be done?

- Fix it
  - Large Data Sets.
  - Train many networks: Ensembles.
  - Train with Outliers

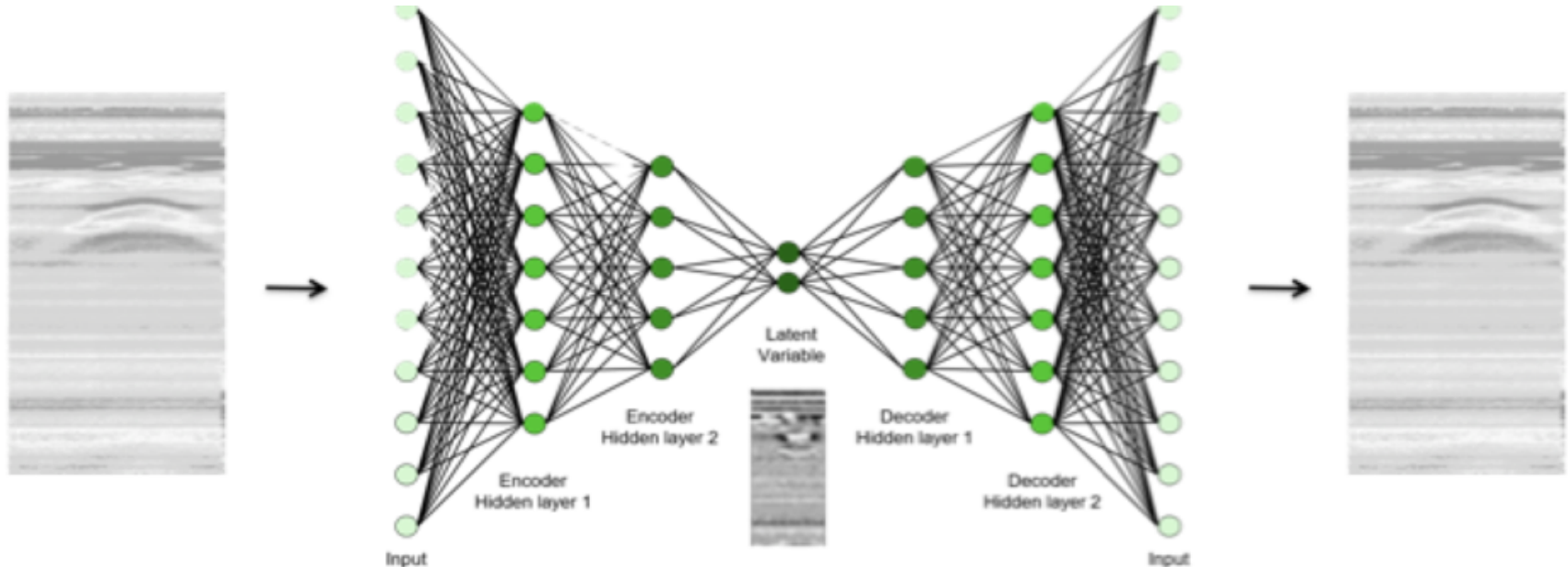




# Auto-Encoders

## Target Output = Input

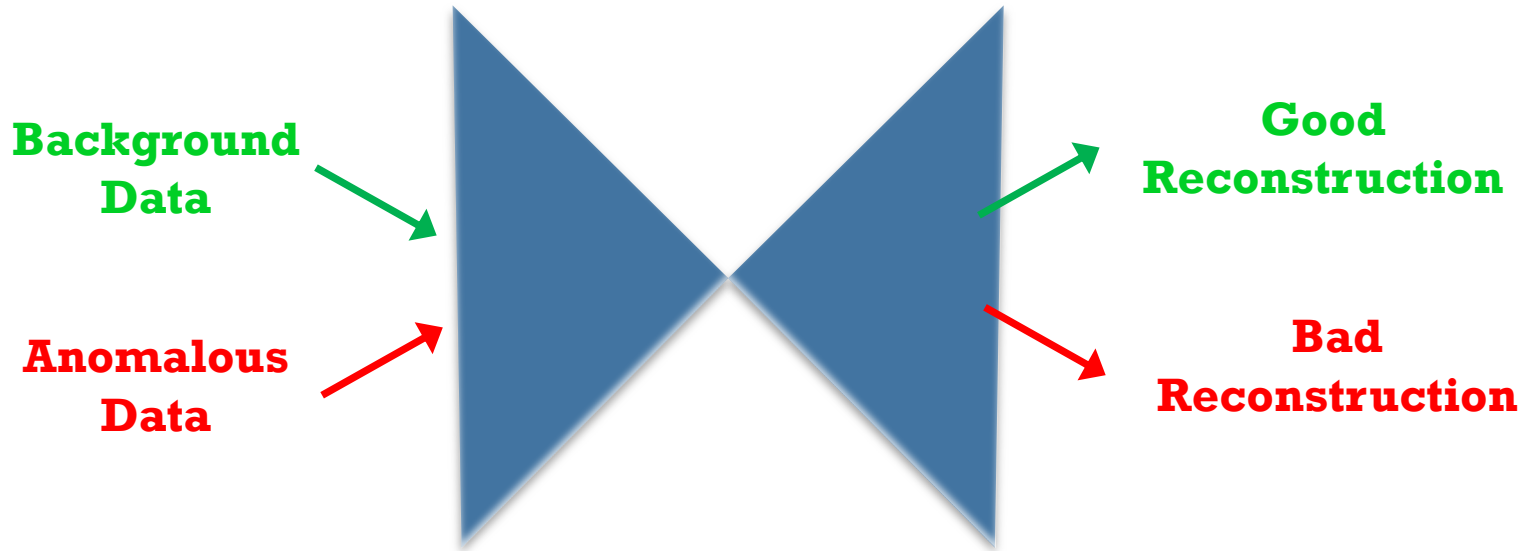
**Reconstructs training data well but not other data (sometimes)**  
Good for Anomaly Detection





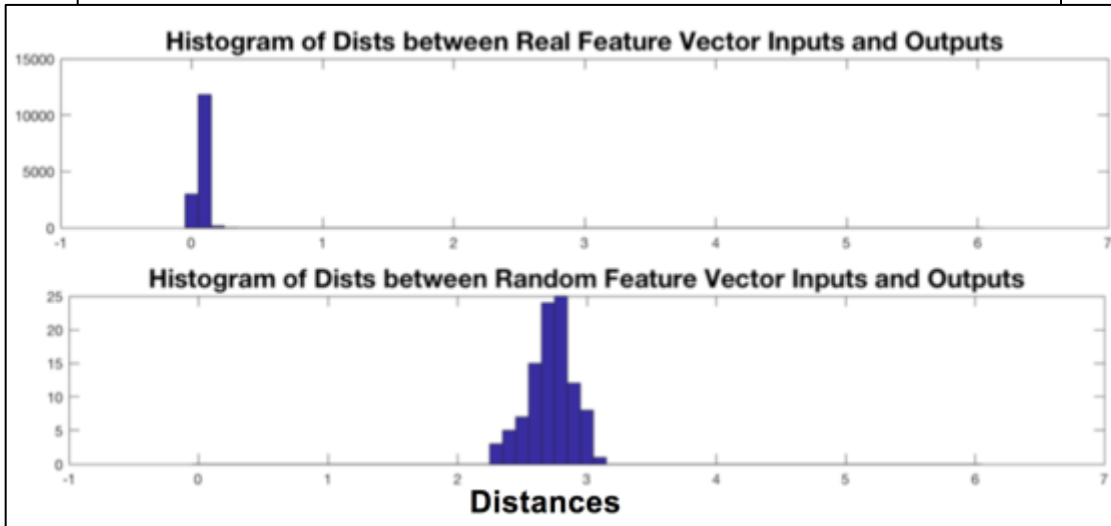
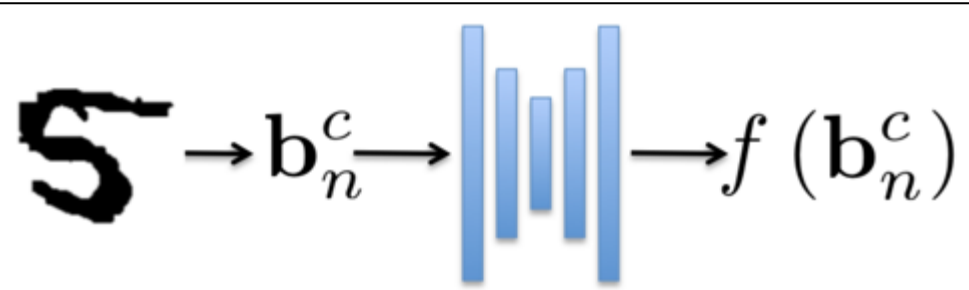
## Auto-Encoders

**Learns to reconstruct training data well but not other data (sometimes)**





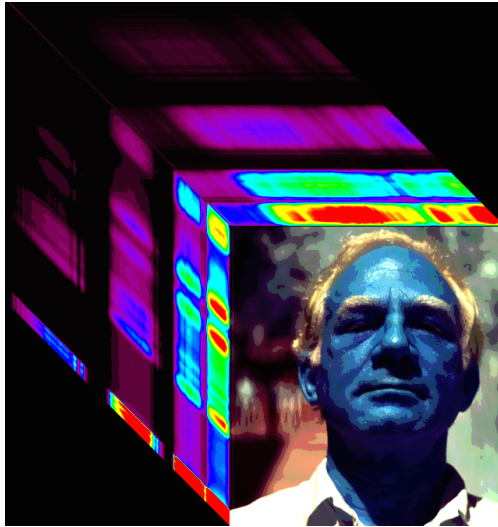
# Auto-Encoders Experiment



- Can construct Anomaly Detectors from background only data
- Learn to reconstruct back

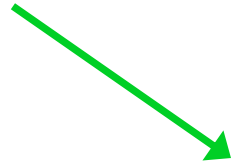


# Keep away from bad places and go to good places

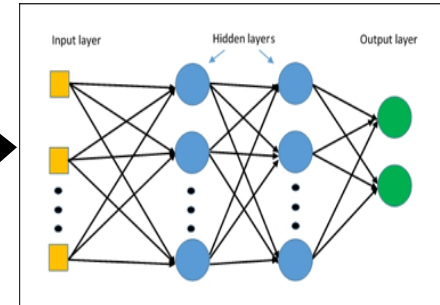
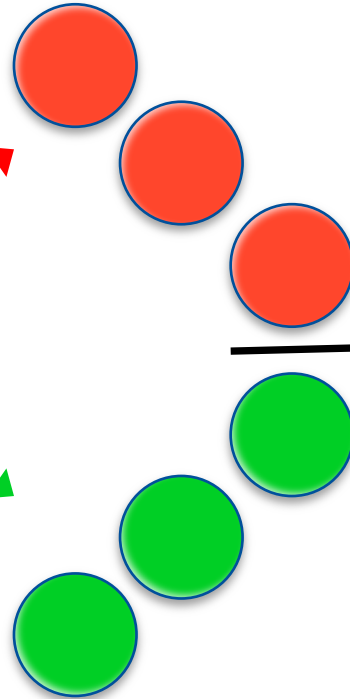


Input Data

**Stay Away**



**Go Here**



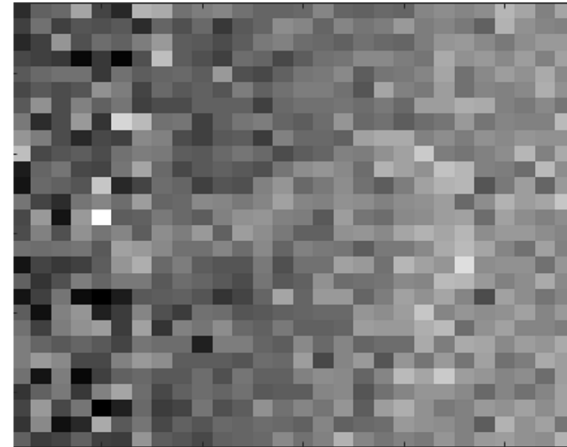


## NULL SPACE IMAGES



Original 2

and



Original 2 + Null Space Image

Produce same output!!



## Self Organizing Map and Manifolds



These patterns represent the  
Good Places.

If you are not close to one of  
them, then you are in a Bad  
Place



## Summary

ANNs are not Robust

Have to be careful

There are several methods for mitigating this problem

Working on some hyperspectral now, results soon