



- Supervised learning
- Nearest neighbor methods
  - What are they?
    - Distance functions.
  - When do they work and when do they not work?
    - Test setting is similar to training setting
    - Images are not overly variable.
  - Theoretical results
- Alternatives?

- Supervised learning:
  - Formalization of the idea of learning from examples.
- 2 elements:
  - Training data
  - Test data
- Training data:
  - Data in which the *class* has been identified.
    - Example: This is a “three”. 
- Test data:
  - Data which the algorithm is supposed to identify.
  - What is this? 



## ■ Formally:

- n training data pairs:

$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_n, y_n)$$

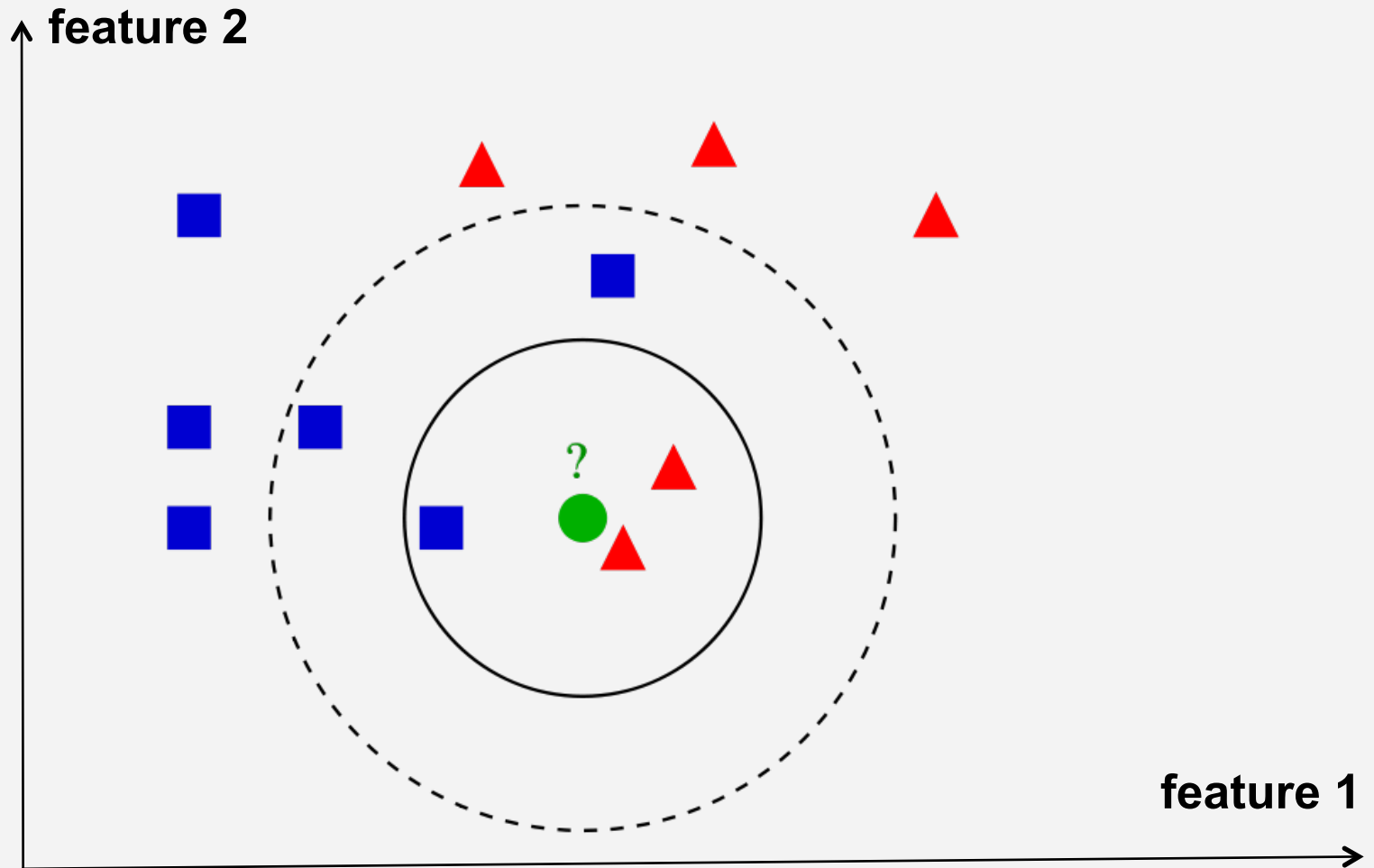
x's are "observations"

y's are the class labels

- m test data samples:

$$(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_{n+m})$$

- Choose label of training example closest to the test example.
- K-nearest neighbor rule (K-NN)
  - Choose some value for  $K$ , often dependent on the amount of data  $N$ .
    - $K = \sqrt{N}$  is a common choice.
    - For a two-class problem,  $K$  is usually odd. (Why?)
  - Among  $K$  nearest neighbors, have a vote for the label.
  - Break ties with a random choice.



$$d(\mathbf{p}, \mathbf{q}) = d(\mathbf{q}, \mathbf{p}) = \sqrt{(q_1 - p_1)^2 + (q_2 - p_2)^2}$$

In two *dimensions*

or

Two *measurements* per point

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# Euclidean distance

$$d(\mathbf{p}, \mathbf{q}) = d(\mathbf{q}, \mathbf{p}) = \sqrt{(q_1 - p_1)^2 + (q_2 - p_2)^2 + \cdots + (q_n - p_n)^2} = \sqrt{\sum_{i=1}^n (q_i - p_i)^2}.$$

In  $n$  *dimensions*

or

$n$  *measurements* per point



```
>> I=rand(5,5)
```

```
I =
```

```
    0.0003    0.5830    0.4411    0.3030    0.5376  
    0.0201    0.5671    0.1130    0.8868    0.6623  
    0.9907    0.4703    0.1832    0.0563    0.4345  
    0.6449    0.9332    0.3108    0.4623    0.4908  
    0.9828    0.9349    0.0914    0.6733    0.8732
```

```
>> J=rand(5,5)
```

```
J =
```

```
    0.7968    0.2026    0.7233    0.1956    0.1599  
    0.9779    0.3315    0.8022    0.5367    0.5875  
    0.0319    0.2626    0.1441    0.5384    0.0207  
    0.5556    0.1637    0.0796    0.4609    0.8187  
    0.1396    0.0544    0.7685    0.1129    0.3389
```

```
>> diff=I(:)-J(:);
```

```
>> diffSquare=diff.^2;
```

```
>> EuclidDist=sqrt(sum(diffSquare));
```

```
^^
```

```
>> EuclidDist=sqrt(sum((I(:)-J(:)).^2))
```

```
EuclidDist =
```

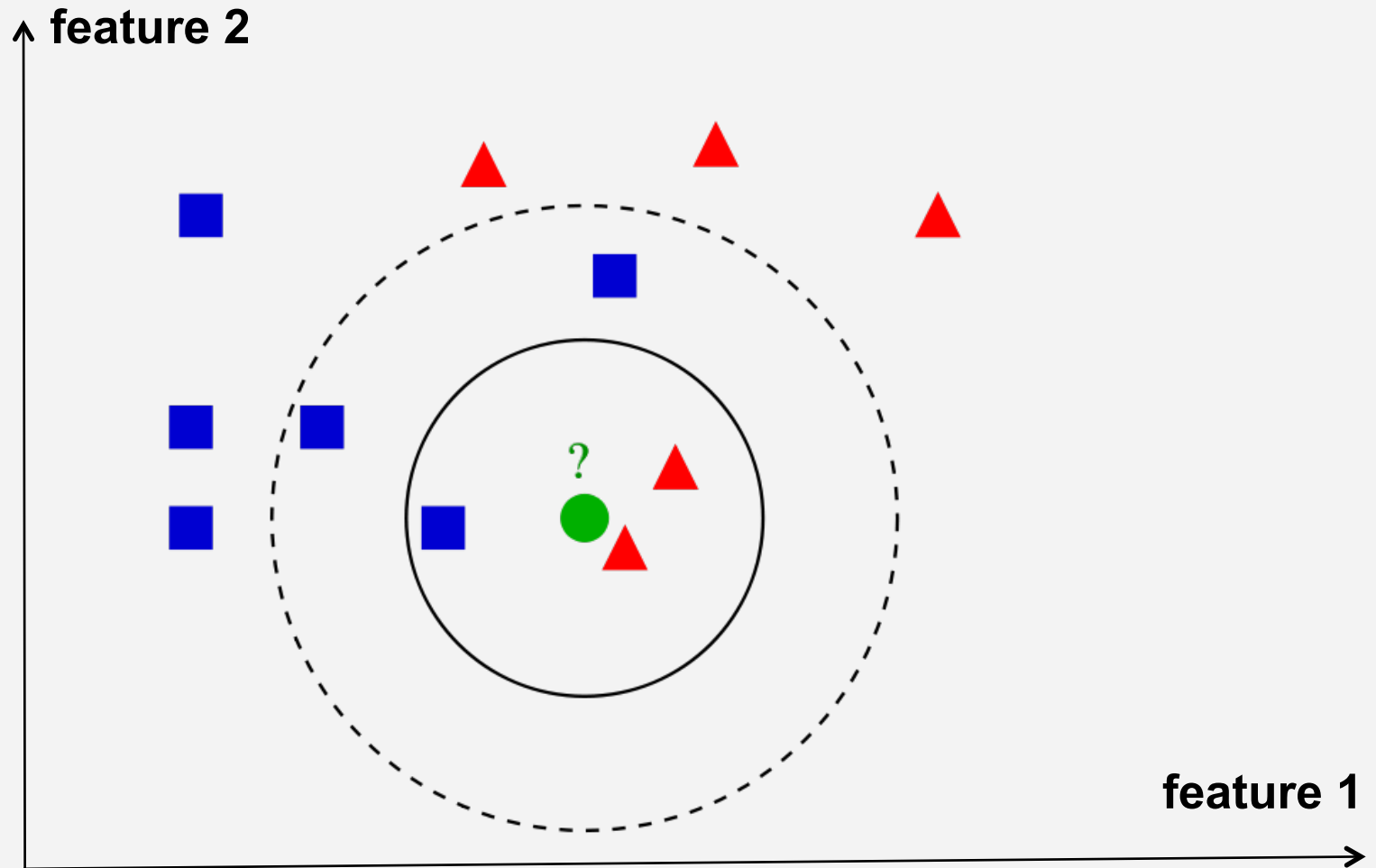
```
2.6947
```

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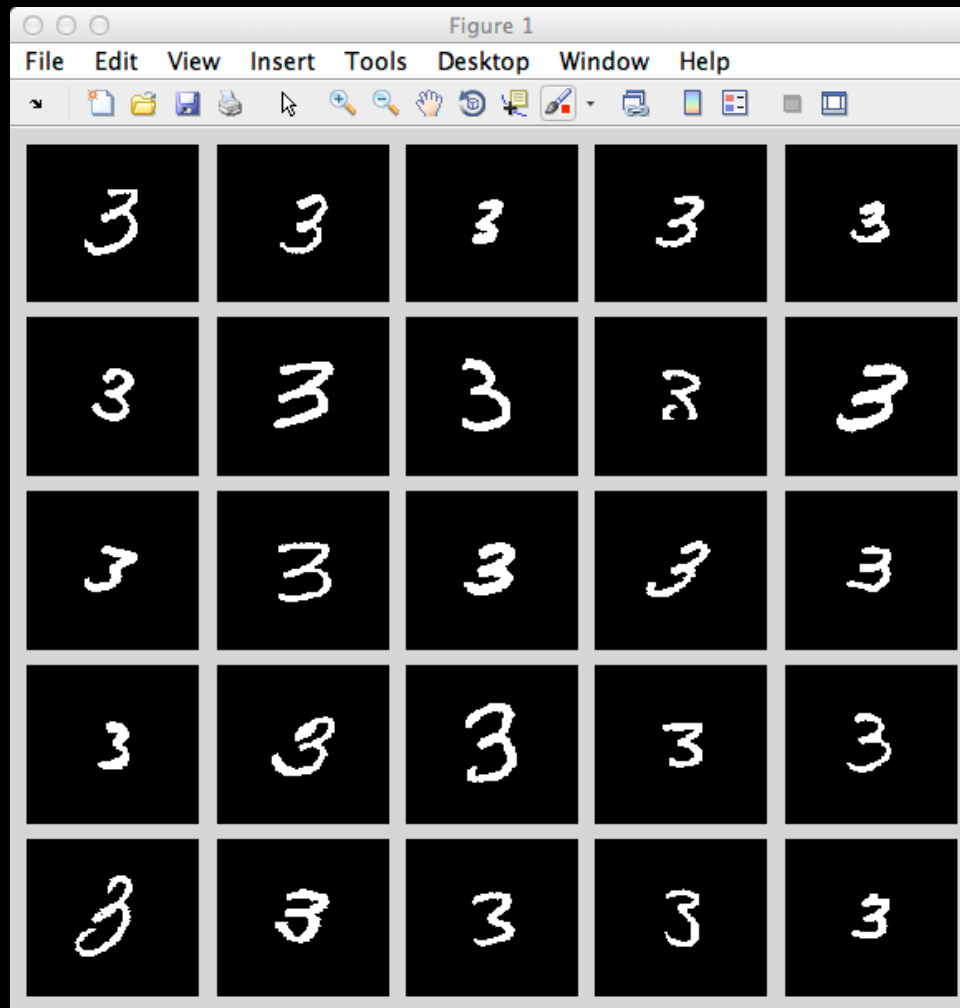
# Nearest Neighbor Rule

- Choose label of training example closest to the test example.

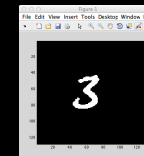


- Binary images: 128x128.
  - How many measurements?
  - How many dimensions?
- 25 training images of 3's
- 25 training images of 5's
- One test example

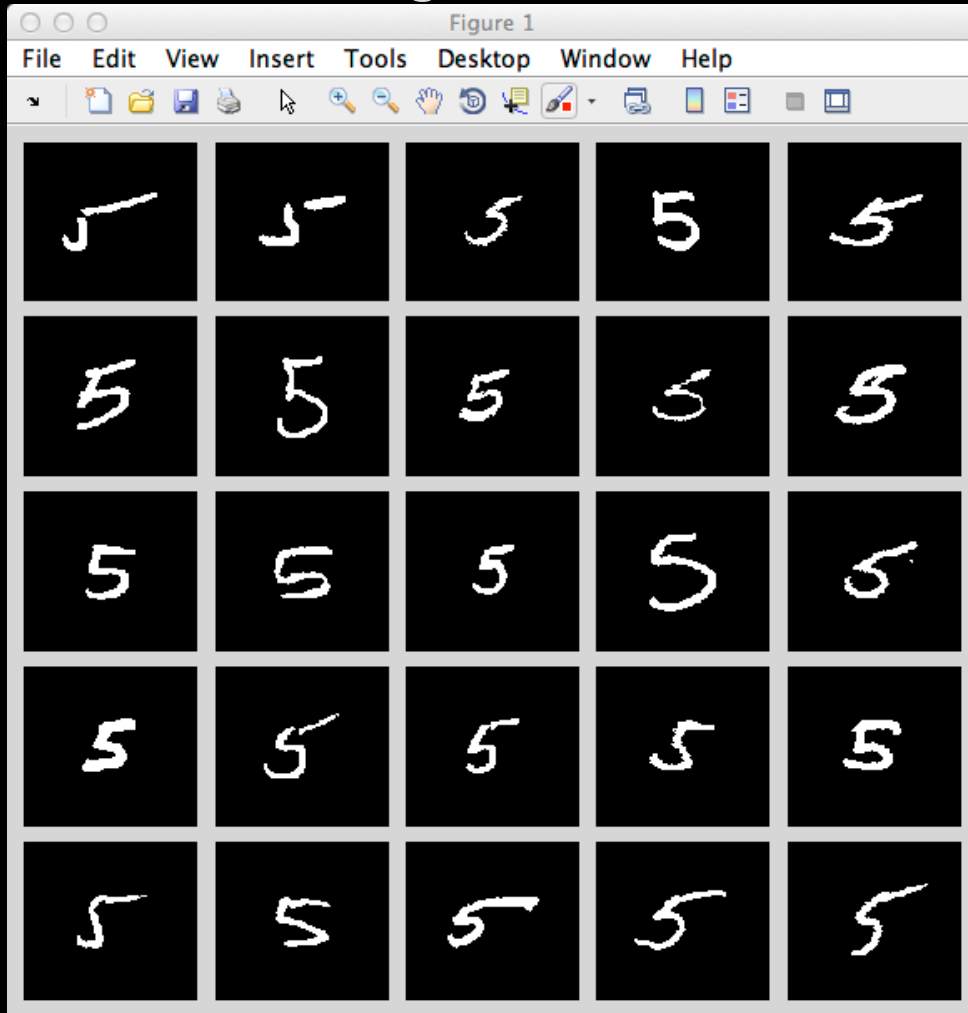
## Training Data for 3's



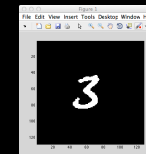
## Test Image



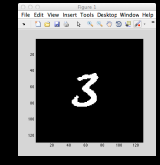
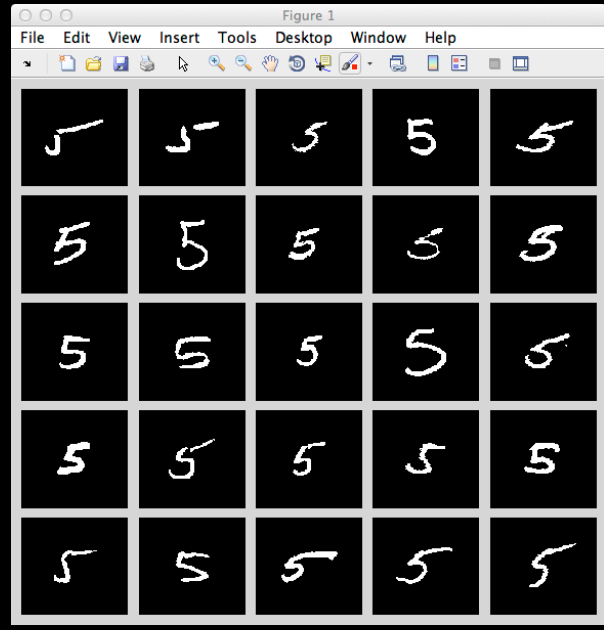
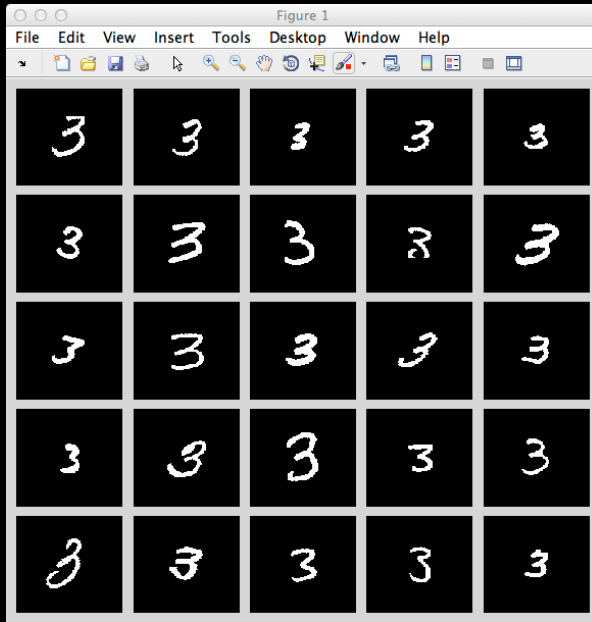
### Training Data for 5's



### Test Image



# What's the Nearest Neighbor?





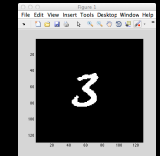
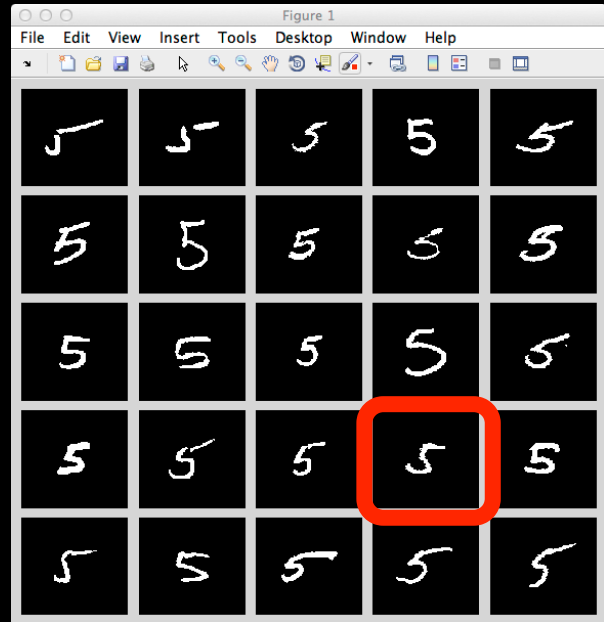
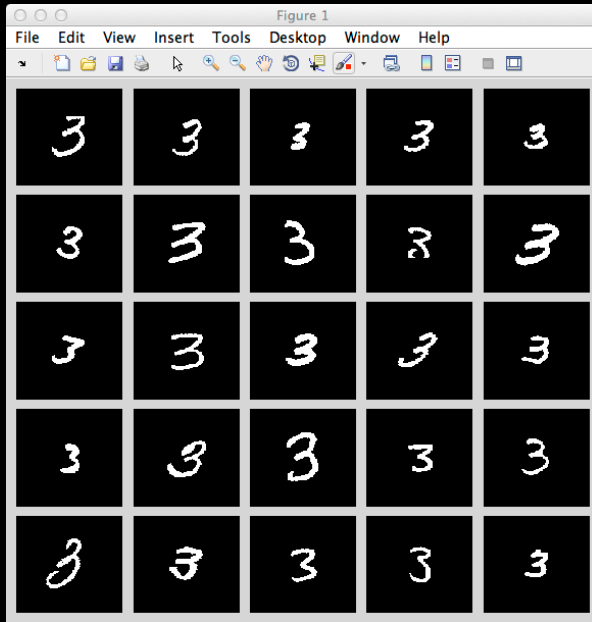


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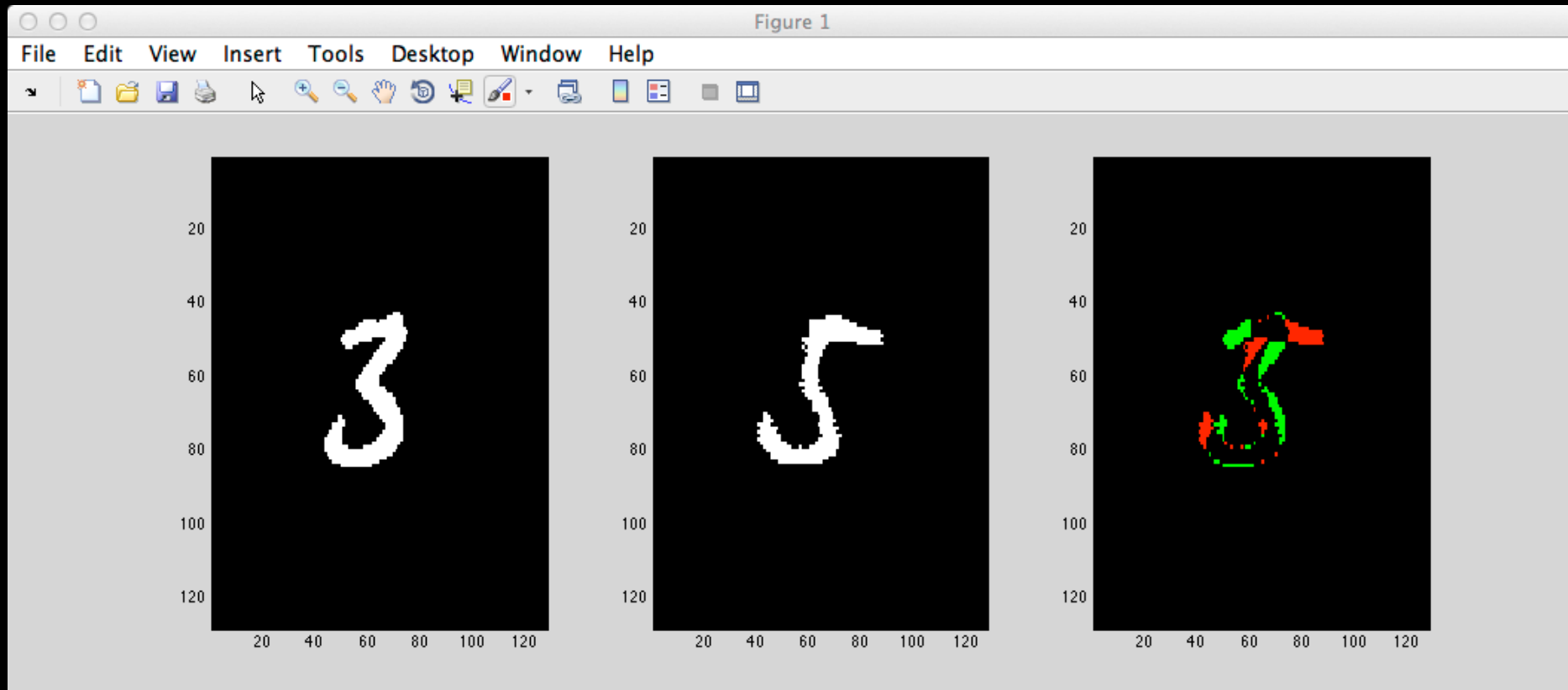
# What's the Nearest Neighbor?



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# Visualizing the image difference



- Treats every pixel the same
- Is thrown off by changes in position, scale, or orientation of image
- Can't handle changes in color