Sudoku

- Rules

- Example Problem (4x4)

- Feasibility Problem

- LP Relaxation

Here is the puzzle. Good luck!

Evil Puzzle for Oct 13, 2012

9x 9

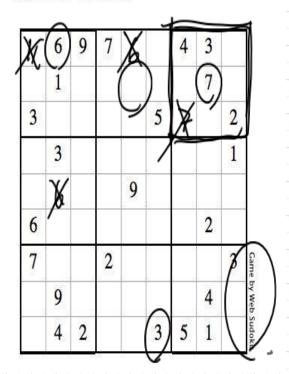
	6	9	7			4	3	
	1						7	
3					5			2
	3		02	`				1
				9				
6							2	
7			2					3
	9						4	
	4	2			3	5	1	

9x 9

- Given a set of "clues"
- Each space takes an Integer between 1-9
- -Each Integer should only show up once
 - -- 1.) in each 3x3 cell
 - -- 2.) in each row
 - -- 3.) in each column
 - -- 4.) in each space
 - -- 5.) No Clue should be replaced
- Need to find a feasible Integer solution

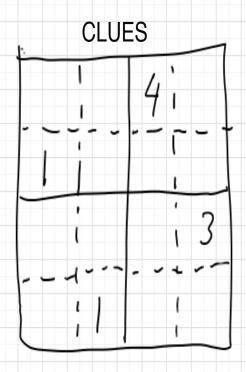
Here is the puzzle. Good luck!

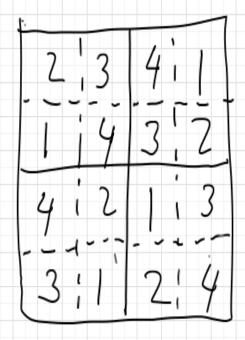
Evil Puzzle for Oct 13, 2012



Simple Example (4x4)

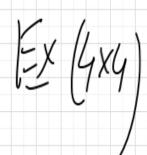
- Each space takes an Integer between 1-4
- Each Integer should only show up once
 - -- 1.) in each 2x2 cell
 - -- 2.) in each row
 - -- 3.) in each column
 - -- 4.) in each space
 - -- 5.) No Clue should be replaced
- Need to find a feasible Integer solution





Feasibility Problem

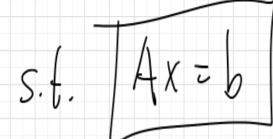
- Integer Programming Problem
 - 21,2,3,--93
 - -- Binary Programming Problem
- How do we define (A) and (b)
 - -- based on the "rules"
 - -- based on the definition of \sqrt{x}



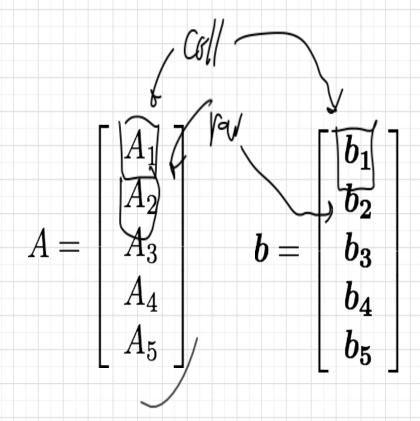
minimize ?subject to Ax = b

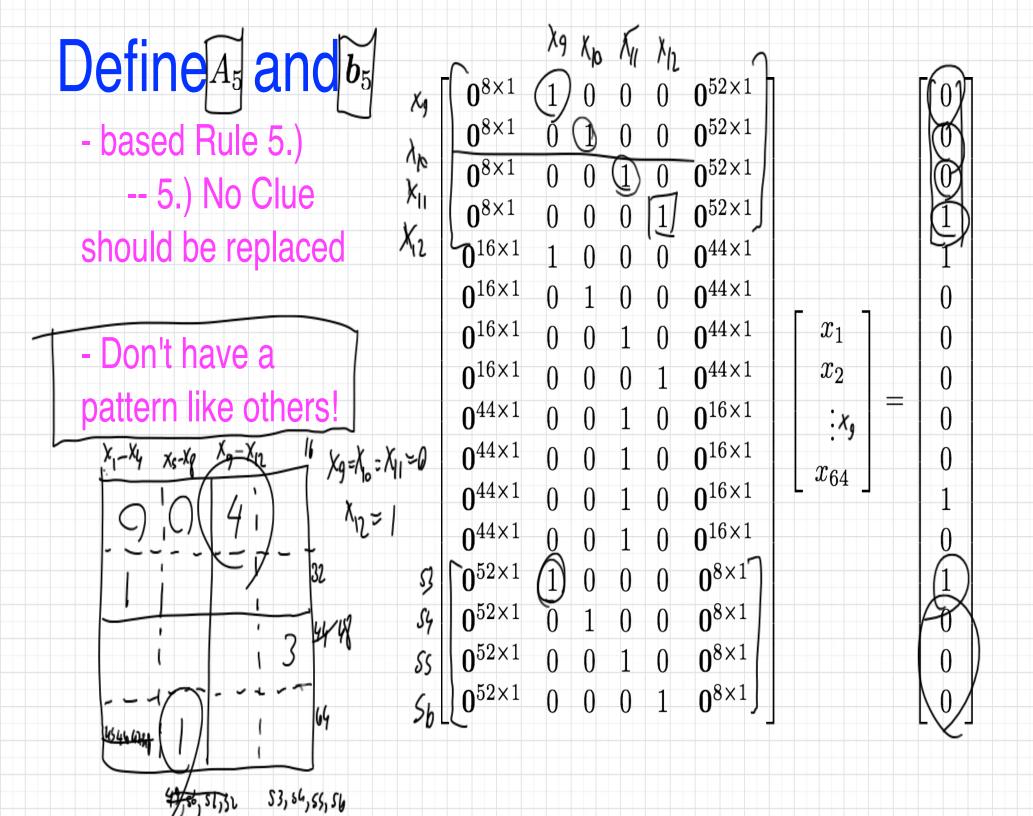
Define the variables $oldsymbol{x}$ space (1,1) = 2(17) - Use the 4x4 example x_3 (1,2) - each space is defined by 4 space (1,2) = 3binary variables x_{10} space (1,3) = 4 x_{11} - A total of $4x4x4 \neq 64$ x =variables x_{12} x_{13} x_{14} space (1,4) = 1 x_{15} x_{16} (1,1) (4,4) x_{61} 59463 x_{62} space (4,4) = 4 x_{63} x_{64}

Define A and b



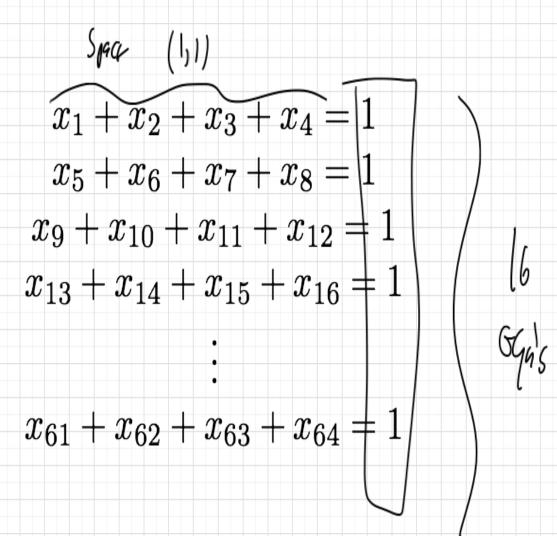
- based on the 5 rules!
- Each Integer should only show up once
 - -- 1.) in each 2x2 (or 3x3) cell
 - -- 2.) in each row
 - -- 3.) in each column
 - -- 4.) in each space
 - -- 5.) No Clue should be replaced





Define \underline{A}_4 and b_4 b_4 b_4

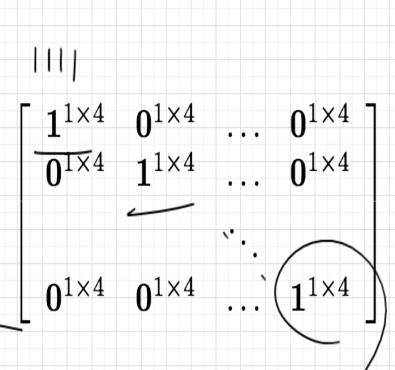
- based Rule 4.)
 -- Each Integer
 should only show
 up once 4.) in each
 space.
- Notice the pattern!

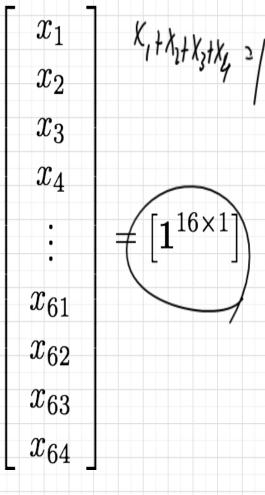


Define A₄ and b₄

- based Rule 4.)
- -- Each Integer should only show up once 4.) in each space.

Notice the pattern!





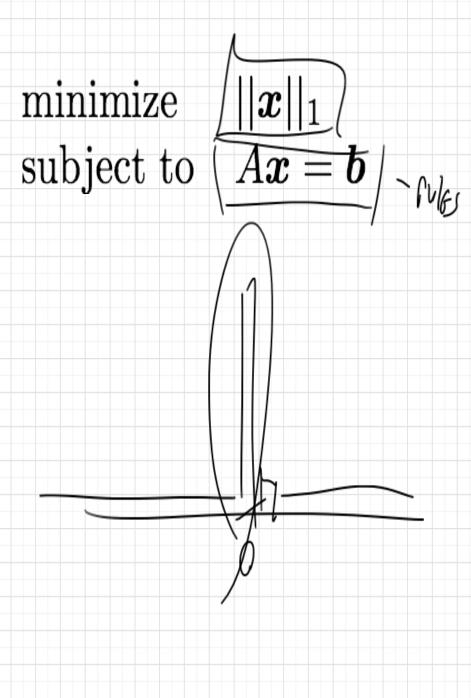
LP Relaxation

- The optimal solution will have many more 0's than 1's ($x_i = 0$)

- RELAX: allow x_i to be any real number between 0 and 1 rather than a binary integer

- Maintain the rules while making as many variable zero

- Remember the Histogram!



Classification (Linear)

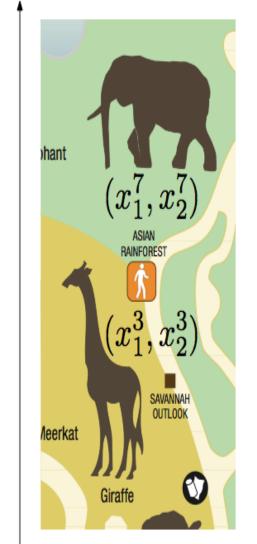
- Figure out, <u>autonomously</u>, which category (or class) an "unknown item" should be categorized into.
- Number of Categories/Classes:
 - * Binary: 2 different categories
 - * Mulitclass: More than 2 categories
- Features: the measurable parts that make up the "unknown item" (or the <u>information</u> you have available to catagorize)

Illustrative Example: The Zoo

- Binary Classification:
 - * Elephants vs Giraffes
- Features:
 - * the coordinate of the "unknown" animal i in the zoo: (x_1^i, x_2^i) or

$$x = \left[egin{array}{c} x_1 \ x_2 \end{array}
ight]$$

 x_2



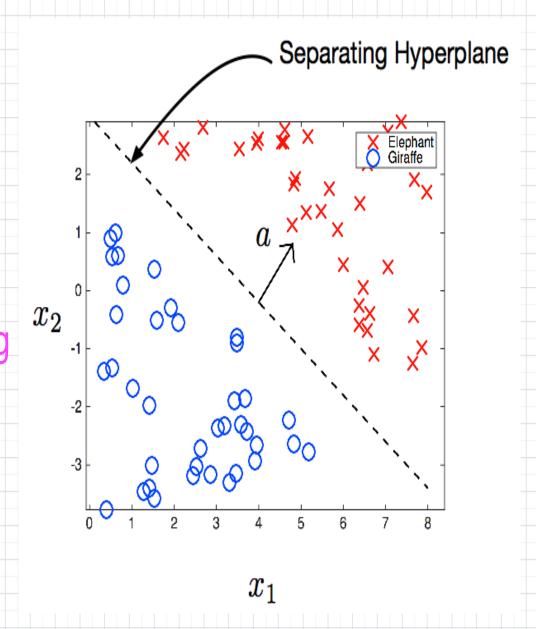
 x_1

The Zoo: cont...

- Is it possible to distinguish between an elephant and a giraffe by it's coordinates on a map of the zoo?
- We need to FIND a separating hyperplane (or a line in 2D):

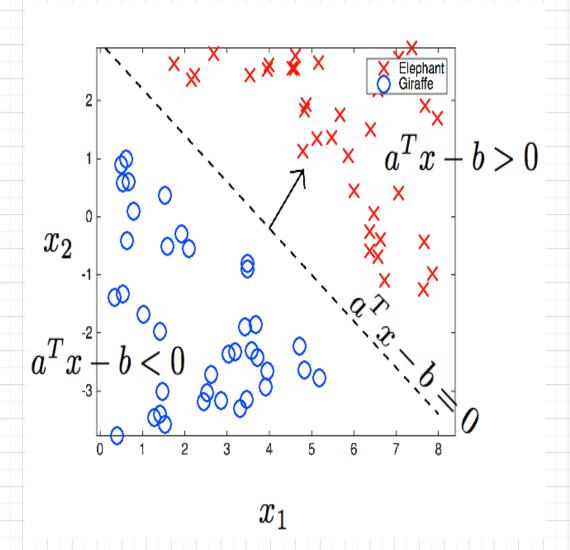
$$\left[\begin{array}{cc} a_1 & a_2 \end{array}\right] \left[\begin{array}{c} x_1 \\ x_2 \end{array}\right] - b = 0$$

or
$$a^T x - b = 0$$



The Zoo: cont...

- Given:
 - * Hyperplane defined by
 - a and b
 - * an animals coordinates (or features) x
- Decision Making:
 - * if $a^Tx b > 0$ than it's an elephant
 - * if $a^Tx b < 0$ than it's a Giraffe



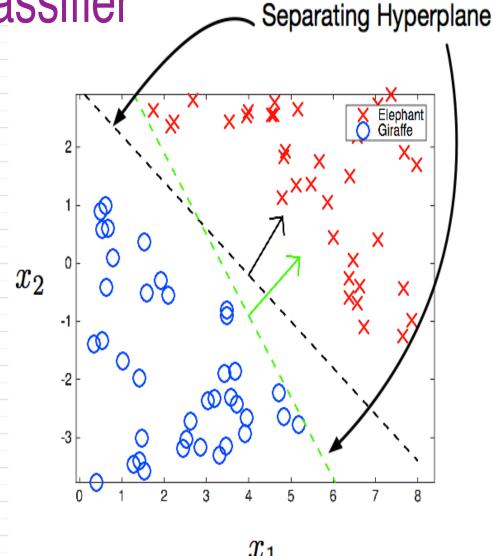
Generate a (Linear) Classifier

- Automate the generation of a hyperplane.(or support vector)

- Training a classifier:

* Given: a "training" set of labeled data:

Example: 100 animals
 (coordinates) are given to
 you, labeled "elephant" or "giraffe"



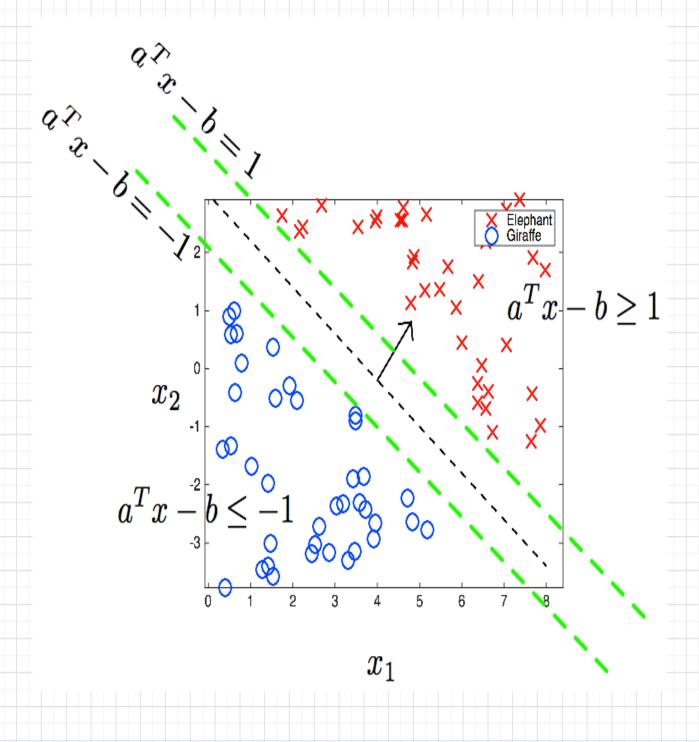
The Zoo: cont...

- Find a and b such that an elephant given $a^Tx - b > 0$

or

- Find a and b such that an elephant given $a^Tx - b \ge 1$

 Same problem if strictly seperable



LP Formulation 1

- assume *n* features (2 for the zoo problem)
- assume m data points $x_i = [x_1^i \ x_2^i]^T$ in training set (so 100 animals)
- assume q (of m) are elephants in the training set
- assume r (of m) are Giraffes
- a and b are the variables (unknown)

minimize Something

subject to

elephant
$$\begin{cases} a^T x^1 - b \ge 1 \\ a^T x^2 - b \ge 1 \\ \vdots \\ a^T x^q - b \ge 1 \end{cases}$$

$$\operatorname{giraffe} \left\{ \begin{array}{l} a^T x^{q+1} - b \leq -1 \\ a^T x^{q+2} - b \leq -1 \\ \vdots \\ a^T x^{q+r} - b \leq -1 \end{array} \right.$$

LP Formulation 2

- assume *n* features (2 for the zoo problem)

minimize
$$\sum_{i=1}^{q} u_i + \sum_{i=1}^{r} v_i$$

subject to

assume m data points in training set (so 100 animals)

- assume q (of m) are elephants in the training set

- assume r (of m) are Giraffes

- need slack variable u and v, where all the elements are positive (u_i , $v_i >= 0$)

$$a^T x^1 - b \ge 1 - u_1$$

 $a^T x^2 - b \ge 1 - u_2$

:
$$a^T x^q - b \ge 1 - u_q$$

$$a^T x^{q+1} - b \le -(1 - v_1)$$

 $a^T x^{q+2} - b \le -(1 - v_2)$

:
$$a^T x^{q+r} - b \le -(1 - v_{q+r})$$

LP Formulation 3

- since $a^Tx - b$ is a scalar, it can also be written as

$$x^Ta-b$$

minimize
$$\mathbf{1}^T u + \mathbf{1}^T v$$

subject to
$$X_ea - b \ge 1 - u$$

$$X_ga - b \le -(1 - v)$$

$$u > 0$$

$$v > 0$$

- Where:

$$X_e = \left[egin{array}{cccc} x_1^1 & x_2^1 & \cdots & x_n^1 \ x_1^2 & x_2^2 & \cdots & x_n^2 \ dots & dots & dots & dots \ x_1^q & x_2^q & \cdots & x_n^q \ \end{array}
ight] = \left[egin{array}{cccc} (x^1)^T \ (x^2)^T \ dots \ dots \ (x^q)^T \ \end{array}
ight] \qquad X_g = \left[egin{array}{cccc} (x^{q+1})^T \ (x^{q+2})^T \ dots \ dots \ (x^{q+r})^T \ \end{array}
ight]$$

and

$$X_g = \left[egin{array}{c} (x^{q+1})^T \ (x^{q+2})^T \ dots \ (x^{q+r})^T \end{array}
ight]$$

Code (using cvx)

```
%%%%%%% Pull out the Giraffes and
%%%%%%% Elephants from the DATA
Giraffes = Animals(Flag2,:);
Elephants = Animals(Flag1,:);
R = length(Elephants(:,1));
Q = length(Giraffes(:,1));
n = 2;
%%%%%%%% Solve for the support vector
cvx_begin
    variables a(n) b(1) u(R) v(Q)
    minimize (ones(1,R)*u + ones(1,Q)*v)
    Elephants*a - b >= 1 - u;
    Giraffes*a - b <= -(1 - v);
    u >= 0;
    v >= 0;
cvx_end
```

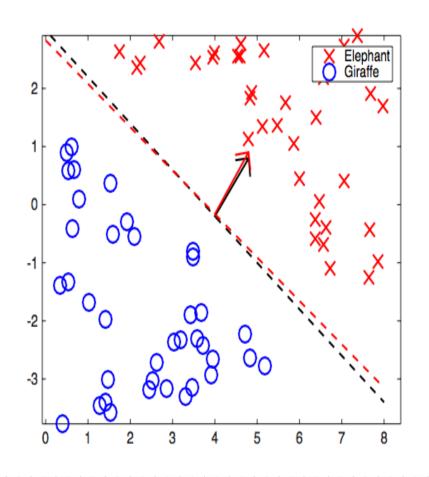
minimize $\mathbf{1}^T u + \mathbf{1}^T v$

subject to
$$X_ea-b\geq \mathbf{1}-u$$
 $X_ga-b\leq -(\mathbf{1}-v)$
 $u>\mathbf{0}$
 $v>\mathbf{0}$

>> Elephants	Solution:	>> a
Elephants =		a =
2.6738 2.8	3013	0.5117 0.6203
Section of the sectio	0902	0.6203
	3566	>> b
	5654 3400	b =
	0614	1.9312
		1.9312

Results 1

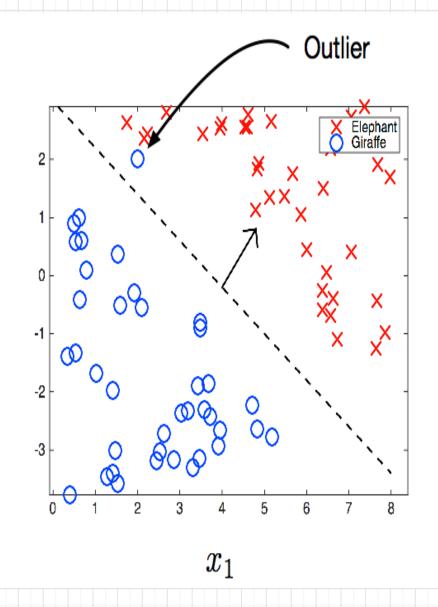
- Black Hyperplane: the actual hyperplane use to produce the x's and o's
- Red Hyperplane: the result of the optimization problem using the x's and o's as training data



Results 2: Outlier

Note that in the "real-world", you may have noise, errors, or outliers that don't accurately represent the actual phenomena.

 x_2



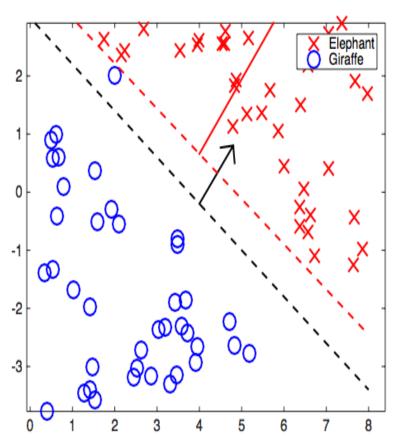
Results 3: Outlier

to the outlier

Notice that the Red Hyperplane,
 is not as accurately
 represent the division due

Q: Can We do better when we have noisy data or outliers'

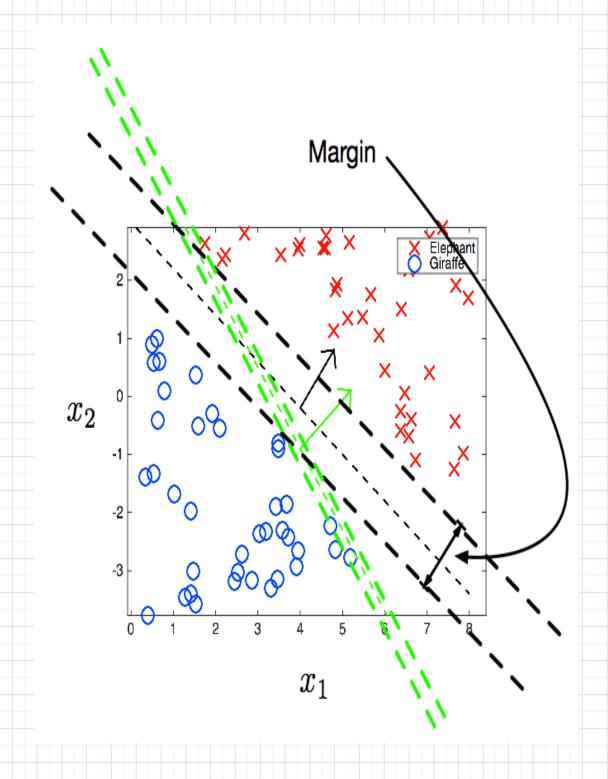
- A: Yes, but we need to look beyond LP.



Maximize Margin

- Black's margin is bigger than green's
- Black is more accurate
- Margin = $\frac{2}{||a||_2}$

- minimize $||a||_2$ to maximize the margin



Code (using cvx)

```
%%%%%%% Pull out the Giraffes and
%%%%%%% Elephants from the DATA
Giraffes = Animals(Flag2,:);
Elephants = Animals(Flag1,:);
R = length(Elephants(:,1));
Q = length(Giraffes(:,1));
n = 2;
g = .1;
%%%%%%%% Solve for the support vector
cvx_begin
   variables a(n) b(1) u(R) v(Q)
    minimize (norm(a) + g*(ones(1,R)*u + ones(1,Q)*v))
   Elephants*a - b >= 1 - u;
   Giraffes*a - b <= -(1 - v);
   u >= 0;
   v >= 0:
cvx end
```

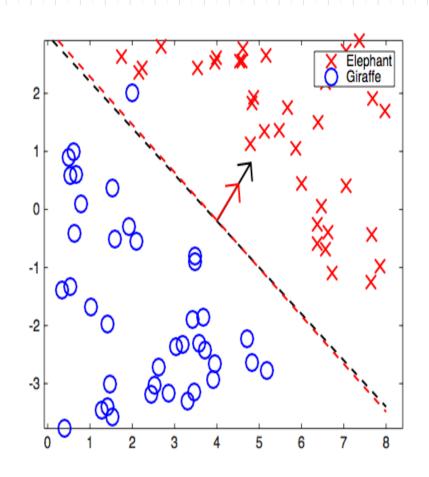
minimize
$$||a||_2 + \gamma (\mathbf{1}^T u + \mathbf{1}^T v)$$

subject to
$$X_ea-b \geq \mathbf{1}-u$$
 $X_ga-b \leq -(\mathbf{1}-v)$
 $u > \mathbf{0}$
 $v > \mathbf{0}$

- use γ as a weighting
 between the following 2
 desires:
 - * Bigger Margin given robustness to outliers
 - * A hyperplane that has few (or no) errors

Results 4: Robust to Outliers

- Notice that the Red Hyperplane,
 is now more representative of
 the true division, but it will have
 an error due to the outlier.
- It is an iterative process to choose a γ that seems to give the "best results"
- Q: What do you mean by "best results"?



Steps to generating a Classifier

- 1.) Split your labelled data into:
 - a.) Training Set
 - b.) Test Set
- 2.) Train your Classifier (pick a γ) using your Training Set
- 3.) Test your Classifier with your Test Set which was not used in the training of the Classifier.
- 4.) Measure accuracy (total error for example) and go back to 2.)
- 5.) Possibly identify a minimal set of feature that has the same predictive power which reduces the model size
- 6.) Use the Classifier

APPLICATION: Handwriting Character Recognition

- United States Postal Service (USPS) uses this to automate zip code reading.
- Multiclass Problem: need to differentiate between digits 0-9
- Q: what is the feature vector and how long is it?

Handwriting Images (Data)

- MNIST Database: yann.lecun.com/exdb/mnist/
- 60,000 character "training" data set and 10,000 character "testing" set
- Each image is a 28 X 28 pixel image where each image is a feature. This gives 784 features! (so n = 784, not 2 like the zoo)



Handwriting Images 2

- Database we will supply you gives you images as 784 long vectors. (ready for optimization)
- To view images, you need to reshape the vector into a28 X 28 matrix of pixels:

```
ImageNumber = 42; % any number between 1-60000
IMAGE = reshape(images(ImageNumber,:),28,28)'
figure,imagesc(IMAGE)
colormap(flipud(gray(256)))
axis equal
set(gca, 'YTick', []);
set(gca, 'XTick', []);
axis off
```



Multiclass Classifier

- Option 1: ONE vs. ALL classifiers (10 total):
 - * 0 vs All the rest
 - * 1 vs All the rest
 - * etc ...
 - * 9 vs All the rest
- Option 2: Pairwise (many BINARY classifiers, 45 total):
 - * 0 vs. 3
 - * 1 vs. 3
 - * 2 vs. 3
 - *
 - * 9 vs 3

Directed - Acyclic - Graph (DAG)

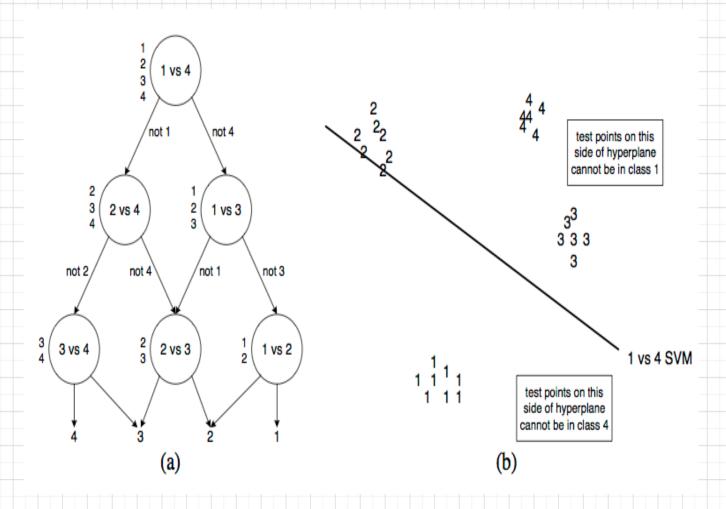
Large Margin DAGs for Multiclass Classification

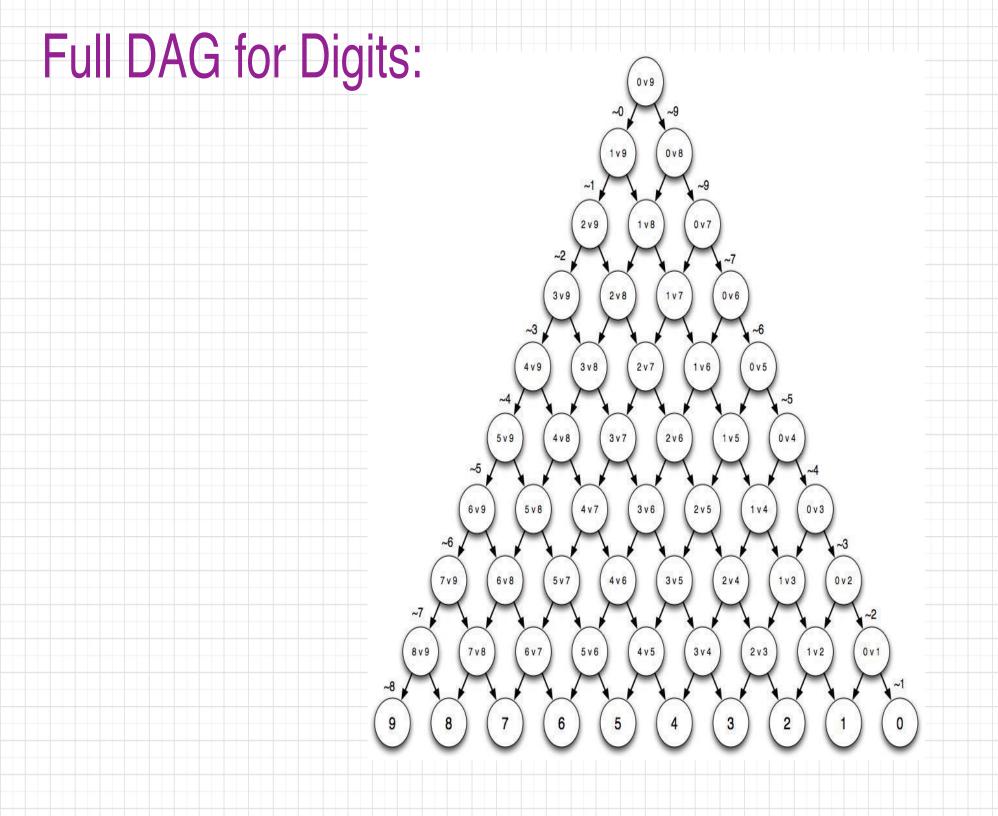
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Steps to create Classifiers for Digit Recognition

- 1.) Train the 45 Binary Classifiers (so 45 different a's and b's)
 - * This will take the most time!
 - * Use the 60,000 training
- images images and their labels
- 3.) Test your Classifier via the DAG
 - * Use the 10,000 testing images
- 4.) Measure accuracy (total error for example) and go back to 2.)

```
%%%%%%% Pull out the zeros and ones from the DATA
Zeros = images(Zeros_Labels,:);
Nines = images(Nines_Labels,:);
R = length(Zeros(:,1));
Q = length(Nines(:,1));
g = .1;
n = 784;
cvx_solver sedumi
%%%%%%% Solve for the support vector
cvx_begin
    cvx_precision low % speed up solver
    variables a(n) b(1) u(R) v(Q)
    minimize (norm(a) + g*(ones(1,R)*u + ones(1,Q)*v))
    double(Zeros)*a - b >= 1 - u;
    double(Nines)*a - b <= -(1 - v);
   u >= 0;
    v >= 0;
cvx_end
toc
%%%%%%% First step of the DAG
Not9 = find(double(images_test)*a - b>0);
Not0 = find(double(images_test)*a - b<0);
```

Results:

Percent_Error_Total =

0.9449

ans =

0.9718
0.9834
0.9513
0.9151
0.9298
0.8975
0.9515

0.9508

0.9404

0.9489

Mistaken as 0:

Mistaken as 9:



4 4 7 4