# Focused Crawling with Scalable Ordinal Regression Solvers

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ICML-2007

# Focused Crawling & Large scale OR

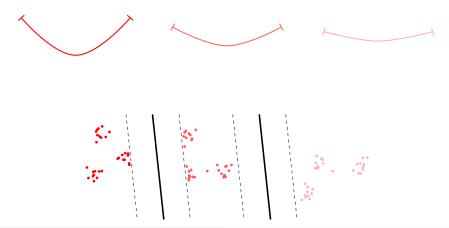
#### Focused Crawling

- Given a topic (seed pages) find out relevant pages from the web
- Pose Focused Crawling as a large scale OR problem

#### Ordinal Regression

- Fast OR training algorithm scales to millions of datapoints
  - Fast algorithm to solve an SOCP with one SOC constraint
- Low prediction time

# Baseline OR Formulation [Chu & Keerthi, 2005]



Describe data using clusters instead of data points

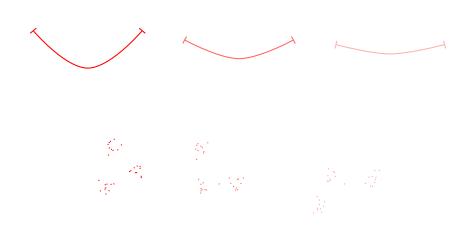
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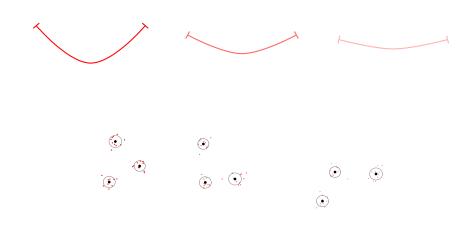
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- Size of optimization problem O(clusters) rather than O(datapoints)

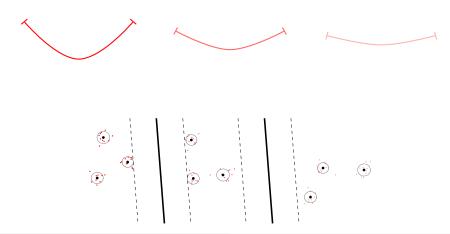
# Proposed OR formulation's solution



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# Proposed OR formulation

#### Features:

- SOCP Problem with one SOC constraint
- $T_{train} = T_{clust} + T_{SOCP} = O(n)$ 
  - Cluster moments estimated using BIRCH [Zhang et.al., 1996]  $T_{clust} = O(n)$
  - SOCP solved using SeDuMi<sup>a</sup>.  $T_{SOCP}$  is independent of n
- Can be Kernelized using input space cluster moments
  - ullet No. of Support Vectors at max. k low prediction time

ahttp://sedumi.mcmaster.ca/

# Clustering + SOCP gives speedup

Table: Training times (sec) with SeDuMi and SMO-OR [Chu & Keerthi, 2005] on synthetic dataset.

	S-Rate	S-Size	SMO-OR	SeDuMi
П	0.002	10,000	182	1
li	0.0025	12,500	260	1
li	0.003	15,000	340	1
li	0.3	1,500,000	×	9
Ш	1	5,000,000	×	36

Table: Training times (sec), test error rate with SeDuMi and SMO-OR [Chu & Keerthi, 2005] on CS-Census dataset.

	S-Size	SMO-OR	SeDuMi
		sec (err)	sec
	5,690	893 (.128)	20.4 (.109)
	11,393	5281.6 (.107)	108.8 (.112)
CS	15,191	9997.5 (.107)	271.1 (.108)
	22,331	×	<b>435.7</b> (.119)

# Large number of clusters is still challenging

Table: Training times (sec), test error rate with SeDuMi and SMO-OR [Chu & Keerthi, 2005] on CH-California Housing dataset.

	S-Size	SMO-OR	SeDuMi
		sec (err)	sec
	10,320	551.9 (.619)	112 (.623)
	13,762	1033.2 (.616)	768.8 (.634)
CH	15,482	1142 (.617)	×
	17,202	1410 (.617)	×
	20,230	1838.5 (.62)	×

#### Key Idea:

- Exploit special SOCP form SOCP problem with one SOC constraint
  - Erdougan et.al., 2006 specialized solvers scale better
- Fast algorithm similar in spirit to Platt's SMO for QP

#### Features:

- More scalable than generic solvers
- Easy to implement, uses no optimization tools

#### Rewrite Dual as follows:

$$\min_{\substack{\alpha,\alpha^* \\ \text{s.t.}}} W\sqrt{(\alpha^* - \alpha)^\top \mathbf{K}(\alpha^* - \alpha)} - \mathbf{d}^\top (\alpha + \alpha^*)$$
s.t. 
$$0 \le \alpha \le 1, 0 \le \alpha^* \le 1$$

$$s_i^* \le s_i, \ \forall \ i = 1, \dots, r - 2, s_{r-1}^* = s_{r-1}$$

K is Gram matrix for cluster centers

$$s_i = \sum_{k=1}^i \sum_{j=1}^{n_k} \alpha_k^j$$
 and  $s_i^* = \sum_{k=2}^{i+1} \sum_{j=1}^{n_k} \alpha_k^{*j}$ 

#### Minimization wrt. two multipliers

$$\min_{\Delta\alpha} \quad \sqrt{a(\Delta\alpha)^2 + 2b(\Delta\alpha) + c} - e\Delta\alpha$$
 s.t. 
$$lb \leq \Delta\alpha \leq ub$$

#### Has closed form solution:

$$\Delta \alpha = \begin{cases} \frac{e\sqrt{\frac{ac-b^2}{a-e^2}} - b}{a} \\ \frac{-b}{a} \end{bmatrix}_{lb}^{ub} & \text{if } ac - b^2 > 0, a - e^2 > 0 \\ \frac{-b}{a} \end{bmatrix}_{lb}^{ub} & \text{if } ac - b^2 = 0, a - e^2 > 0 \\ ub & \text{if } e - \sqrt{a} \ge 0 \\ lb & \text{if } e + \sqrt{a} \le 0 \end{cases}$$

#### CB-OR Algorithm

- Step 1 Pick two most KKT violators
- Step 2 Solve the 1-d minimization problem
- Step 3 Update unknowns
- Step 4 Check for KKT violators. If none terminate. Else Step 1

#### CB-OR — Evaluation

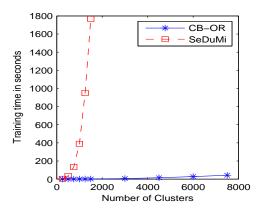


Figure: Dashed line represents training time with **SeDuMi** and continuous line that with **CB-OR** on a synthetic dataset.

#### CB-OR — Evaluation

Table: Comparison of training times (in sec) with **CB-OR**, **SMO-OR** and **SeDuMi** on benchmark datasets. The test set error rate is given in brackets. (CH-California Housing, CS-Census datasets).

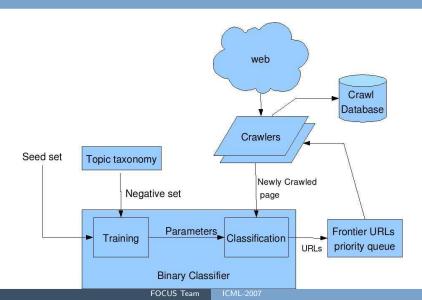
	S-Size	CB-OR	SMO-OR	SeDuMi
		sec (err)	sec (err)	sec
	10,320	.5 (.623)	551.9 (.619)	112
	13,762	1.5 (.634)	1033.2 (.616)	768.8
CH	15,482	8.4 (.618)	1142 (.617)	×
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	20,230	<b>10.4</b> (.62)	1838.5 (.62)	×
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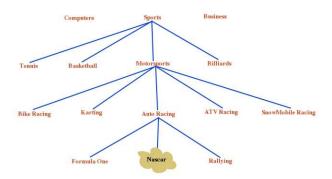
## **Focused Crawling**

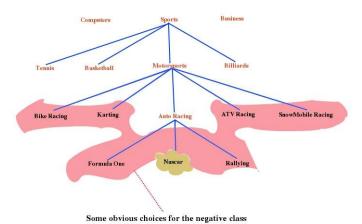
#### Focused Crawling

- Given a topic (seed pages) find out relevant pages from the web.
- S. Chakrabarti et.al (1999,2002), C. Aggarwal et.al (2001), M. Diligenti et.al (2000)
- Requires low bandwidth and low disk space.
- Small updation cycle.

# Baseline Focused Crawler [Chakrabarti et.al., 1999]





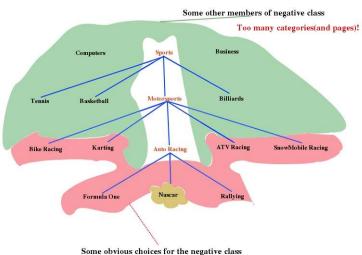


#### Some other members of negative class Business Computers Sports Motorsports Billiards Basketball Tennis ATV Racing SnowMobile Racing **Bike Racing** Karting Auto Racing Nascar Formula One Rallying

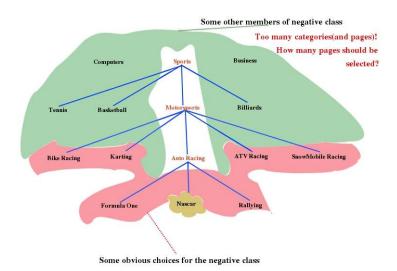
FOCUS Team

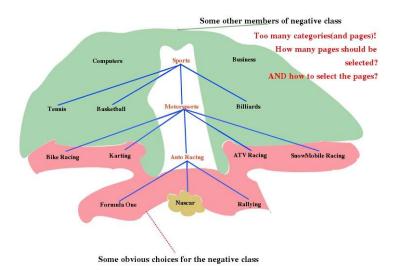
Some obvious choices for the negative class

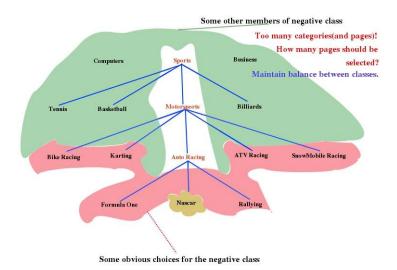
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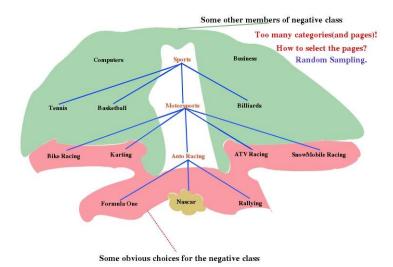


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## Exploit link structure

- Grangier and Bengio observe that hyperlinked documents are semantically closer.
- One link away pages are more similar to seed pages compare to two link away pages.

#### Link structure in web



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# Focused Crawling as OR problem — exploit link structure



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Level 0 - Pages belong to topic

## Focused Crawling as OR problem — exploit link structure



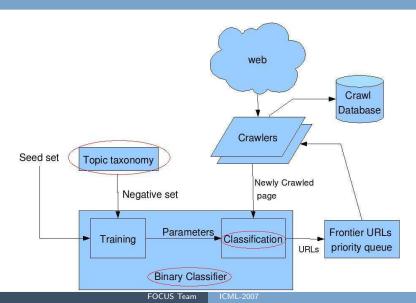
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## Focused Crawling as OR problem — exploit link structure

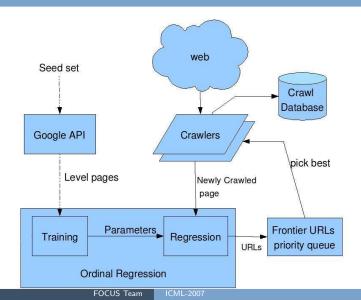


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## Baseline Focused Crawling architecture



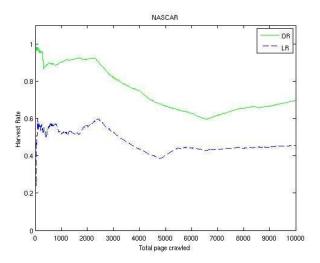
## Proposed Focused Crawling architecture



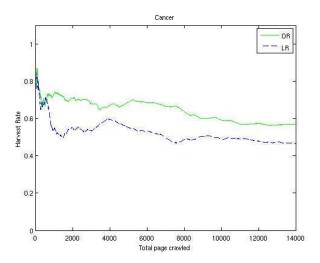
## Focused Crawling is a large scale OR problem

Category	Seed	1	2	3	4
NASCAR	1705	1944	1747	1464	1177
Soccer	119	750	1109	1542	3149
Cancer	138	760	895	858	660
Mutual Funds	371	395	540	813	1059

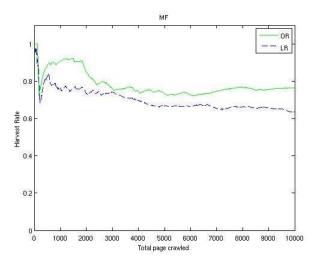
## NASCAR harvest rate



## Cancer harvest rate



## Mutual Funds harvest rate



# Harvest rate comparison

Dataset	Baseline	OR
NASCAR	.3698	.6977
Cancer	.4714	.58
Mutual Fund	.526	.5969
Soccer	.34	.4952

#### Conclusions

- Proposed a scalable clustering based OR formulation
  - Training time O(datapoints)
  - Support Vectors O(clusters)
- Exploited special structure of the formulation to develop a fast solver. CB-OR
  - Scalable to tens of thousands of clusters
- We formulated focused crawling as large scale ordinal regression
  - No need for negative class definition
  - Independent of topic taxonomy
  - OR captures link structure of web graph.

#### Focused crawler code available at

http://mllab.csa.iisc.ernet.in/downloads/focusedcrawler.html

## Acknowledgments

This project is partially supported by AOL India Pvt Ltd and DST, Government Of India (DST/ECA/CB/660)

# Questions?