# Microsoft Research <br> A combinatorial prediction market for the U.S. Elections 

Miroslav Dudík

Thanks: S Lahaie, D Pennock, D Rothschild, D Osherson, A Wang, C Herget


OCTOBER 28, 2012

## CASSIDY'S COUNT: CAN ROMNEY WIN WITHOUT OHIO?

## POSTED BY JOHN CASSIDY



## ELECTORAL COLLEGE: 2012 <br> - leareng obama - lename romety <br>  <br> Mitt Romney's road to presidency this fall looks narrow on electoral map

By Chris Cillizza, April 29, 2012


It's no secret that former Massachusetts governor Mitt Romney has a narrow path to win the presidency this fall. Nowhere is that reality more apparent than when examining the electoral map on which Romney and President Obama will battle in November.
A detailed analysis of Romney's various paths to the 270 electoral votes he would need to claim the presidency suggests he has a ceiling of

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ELECTORAL COLLEGE:


## FiveThirty Fight

Nate Silver's Political Calculus

Sentember 12. 2012 8:13 am | 181 Comments

## Why Romney is losing must-win Ohio

By Peter Hamby, CNN Political Reporter
updated 5:15 PM EDT, Wed September 26, 2012

POLITICS

## DEBATE BUMP PUTS ROMNEY WITHIN A POINT OF OBAMA IN MUST-WIN OHIO

## Romney, Obama hit must-win

## states

One day before Election Day campaigns se


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Virginia remains key to the roadDiscuss
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Mitt Romney's road to presidency this fall looks narrow on electoral map

By Chris Cillizza, April 29, 2012


It's no secret that former Massachusetts governor Mitt Romney has a narrow path to win the presidency this fall. Nowhere is that realitv

## Romney treating Ohio as a must-win state



## CASSIDY'S COUNT: CAN ROMNEY WIN WITHOUT OHIO?

Ohio: Romney 49\%, Obama 49\% (Romney Must Win VA, FL, And Either OH/WI) RasmussenReports ^ | November 05, 2012

Posted on Tuesday, November 06, 2012 1:37:02 AM by Steelfish

## 

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Romney states

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## Total 2012 election spending: $\$ 7$ billion

By Jake Harper | Jan $312013 \mid$ 11:47 a.m.
A new estimate from the Federal Election Commission puts total spending for the 2012 election at more than $\$ 7$ billion - $\$ 1$ billion more than previously thought.

New FEC Chair Ellen Weintraub unveiled the latest estimate of the 2012 campaign's record-shattering cost at the agency's first open meeting of 2013, one that saw the departure of Cynthia Bauerly, one of the three Democratic commissioners. Though campaign spending was expected to break records after the Supreme Court's 2010 Citizens United
 decision that opened the door for unlimited contributions, the latest FEC estimate exceeds earlier expectations.

Ohio

## s fall looks

r Massachusetts pas a narrow path to win Nowhere is that reality

Youte watching:


## Polling

## accurate, but costly

limited range of questions
limited timeliness

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## Prediction markets

accurate and cheap (after fixed cost) broad range of questions good timeliness

## Outline

Prediction markets: Setting and challenges

Addressing the challenges: constraint generation

Empirical evaluation:
U.S. Elections 2008

Field experiment:
U.S. Elections 2012

## Security

= proposition which becomes true or false at some point in future
"Romney will win Florida in Elections 2012"

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= proposition which becomes true or false at some point in future
"Romney will win Florida in Elections 2012"

Traders buy shares for some price: $\$ 0.45$ per share
For each share of a security receive:
\$1 if true
\$0 if false

# Market implementation: (automated) market maker 


market maker
sets prices
if more shares bought, price increases
the price equals the consensus probability of the event

## Combinatorial securities:

## more information

payoff is a function of common variables e.g., $\mathbf{5 0}$ states elect Obama or Romney


## Combinatorial securities:

## more information

Obama to lose FL, but win election
Obama to win >8 of 10
Northeastern states


# Industry standard: ignore relationships 

Treat them as independent markets:
Las Vegas sports betting
Kentucky horse racing
Wall Street stock options
Betfair political betting

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## Problem:

arbitrage opportunities

## Arbitrage

trading with guaranteed profits

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trading with guaranteed profits possible if prices incoherent
prices cannot be realized as probabilities

price $\$ 0.40$

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Pricing without arbitrage: \#P-hard
Industry standard = Ignore arbitrage

## Arbitrage

trading with guaranteed profits possible if prices incoherent
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price $\$ 0.40$

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Pricing without arbitrage: \#P-hard
Industry standard = Ignore arbitrage
$-$ traders rewarded for computation instead of information
poor information sharing

## Our approach:

## partial arbitrage removal

Separate pricing (must be fast) and information propagation

- fast pricing in independent markets for tractably small groups of securities
- in parallel: constraint generation to find and remove arbitrage

Embedded in convex optimization (with many nice properties).

## Cost-based pricing

## (Chen and Pennock 2007)

Setup:
$n$ securities
$C: \mathbb{R}^{n} \rightarrow \mathbb{R}$ convex cost function
$q \in \mathbb{R}^{n}$ market state = \#shares sold

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$\boldsymbol{q}=(\quad 100$,

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cost: $C(q+r)-C(q)$

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$\boldsymbol{r}=(\quad 0$,
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state updated: $q^{\prime} \leftarrow q+r$
$\boldsymbol{q}^{\prime}=\left(\begin{array}{lll}100 & 402\end{array}\right)$
instantaneous prices: $\nabla C(q)$


$$
\boldsymbol{q}=(\quad 100,400)
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## Cost-based pricing

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$$
\boldsymbol{q}=(\quad 100,1400)
$$


$\boldsymbol{q}^{\prime}=(\quad 100,402)$
$\nabla C(\boldsymbol{q})=(\$ 0.70, \quad \$ 0.75)$

# Can we just use existing approaches from graphical models? 

## MCMC—randomized, slow convergence

 mean field—non-convex belief propagation-lack of convergence
# Can we just use existing approaches from graphical models? 

## MCMC—randomized, slow convergence

 mean field—non-convex belief propagation-lack of convergenceProblematic for pricing:
poor convergence $\rightarrow$ volatility
non-determinism $\quad \rightarrow$ distorted incentives
and user experience

## Our approach

implement a coherent pricing scheme on small groups of securities; e.g.,

FL priced $\frac{e^{q_{1}}}{e^{q_{1}}+e^{q_{2}}}$

number of shares
bought so far

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detect incoherence between groups
act as an arbitrageur to restore coherence

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priced $\frac{e^{q_{2}}}{e^{q_{1}}+e^{q_{2}}}$
detect incoherence between groups
act as an arbitrageur to restore coherence
caveat:

- difficult to detect incoherence in general
- we detect only a subset of violations


## For U.S. Elections: conjunction market

create 50 states (groups of size 2) create all pairs of states (groups of size 4)
for conjunctions of 3 or more, group with opposite disjunction:
$A \wedge B \wedge C$ with $\bar{A} \vee \bar{B} \vee \bar{C}$ (groups of size 2)

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$A \wedge B \wedge C$ with $\bar{A} \vee \bar{B} \vee \bar{C}$ (groups of size 2)
each group is independent market: fast pricing
in parallel: generate, find, and fix constraints

## Local coherence

Pairs:

$$
P[A \wedge B]+P[A \wedge \bar{B}]=P[A]
$$

Larger conjunctions:

$$
P\left[A_{1} \wedge A_{2} \wedge \cdots \wedge A_{m}\right] \leq P\left[A_{i}\right]
$$

## Clique constraints

For a disjunction $A_{1} \vee \cdots \vee A_{m}$, pick a subset $A_{i_{1}} \vee \cdots \vee A_{i_{k}}$

$$
P\left[A_{1} \vee \cdots \vee A_{m}\right] \geq P\left[A_{i_{1}} \vee \cdots \vee A_{i_{k}}\right]
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\begin{aligned}
P\left[A_{1} \vee \cdots \vee A_{m}\right] & \geq P\left[A_{i_{1}} \vee \cdots \vee A_{i_{k}}\right] \\
& \geq \sum_{j=1}^{k} P\left[A_{i_{j}}\right]-\sum_{1 \leq j<l \leq k} P\left[A_{i_{j}} \wedge A_{i_{l}}\right]
\end{aligned}
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$$

\#clique constraints exponential
$\rightarrow$ find only the tightest one!
(approximate submodular maximization via Feige et al. 2007)

## Tree constraints

(Galambos and Simoneli 1996)
For a disjunction $A_{1} \vee \cdots \vee A_{m}$,

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## Tree constraints

(Galambos and Simoneli 1996)
For a disjunction $A_{1} \vee \cdots \vee A_{m}$,

$$
P\left[A_{1} \vee \cdots \vee A_{m}\right] \leq \sum_{i=1}^{m} P\left[A_{i}\right]-\sum_{(i, j) \in T} P\left[A_{i} \wedge A_{j}\right]
$$

where $T$ is a spanning tree on nodes $1, \ldots, m$

## Does it work?

Tested using a survey of Election 2008:
singletons, pairs, triples
Small data set-compare with exact:
10 states, 30k trades
Large data set-compare with independent:
50 states, 300k trades

## Small data set: 10 states



Independent LMSR
Local
Clique
Tree
Clique, Tree
$\circ$
$\Delta$
$+$
$\times$
$\diamond$
$\square$ -

## Small data set: 10 states



| Independent | $\circ$ | $\square$ |
| :--- | :--- | :--- |
| LMSR | $\Delta$ | $\square$ |
| Local | + | $\square$ |
| Clique | $\times$ | $\square$ |
| Tree | $\diamond$ | $\square$ |
| Clique, Tree | $\nabla$ | $=$ |

## Large data set: 50 states, 300k trades


sensitivity parameter

## No really, does it work?

# WiseQ Game 

(launched September 16, 2012)

WiseQ Game - Elections 2012 (Beta)

MAKE A PREDICTION

| President | Senate | Governors | Leaderboards | My Portfolio | My Leagues | Forum | FAQ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |

WiseQ Game - Elections 2012 (Beta)


WiseQ Game - Elections 2012 (Beta)


## WiseQ by numbers

437 active users
3,137 trades
514 distinct bundles traded
$10^{33}$ possible outcomes
44.5 million possible bundles allowed by our menu

17,222 securities in 2,840 small markets 20,983 coherence constraints

## Did market absorb information from users?

## Did market absorb information from users?



## Did users place combinatorial bets?

## Did users place combinatorial bets?



## Did users place combinatorial bets?



## Did users place combinatorial bets?



## Did users place combinatorial bets?



## Numerical predictions: electoral votes

## Numerical predictions: electoral votes



## Numerical predictions:

 job numbers

## Summary

independent markets + constraints:
tractable and accurate
combinatorial markets can succeed with
moderate numbers of users
even on huge outcome spaces
meaningful forecasts for
challenging, but relevant outcomes:
combinatorial and numerical

