

# AN OPTIMIZATION-BASED FRAMEWORK FOR AUTOMATED MARKET MAKING

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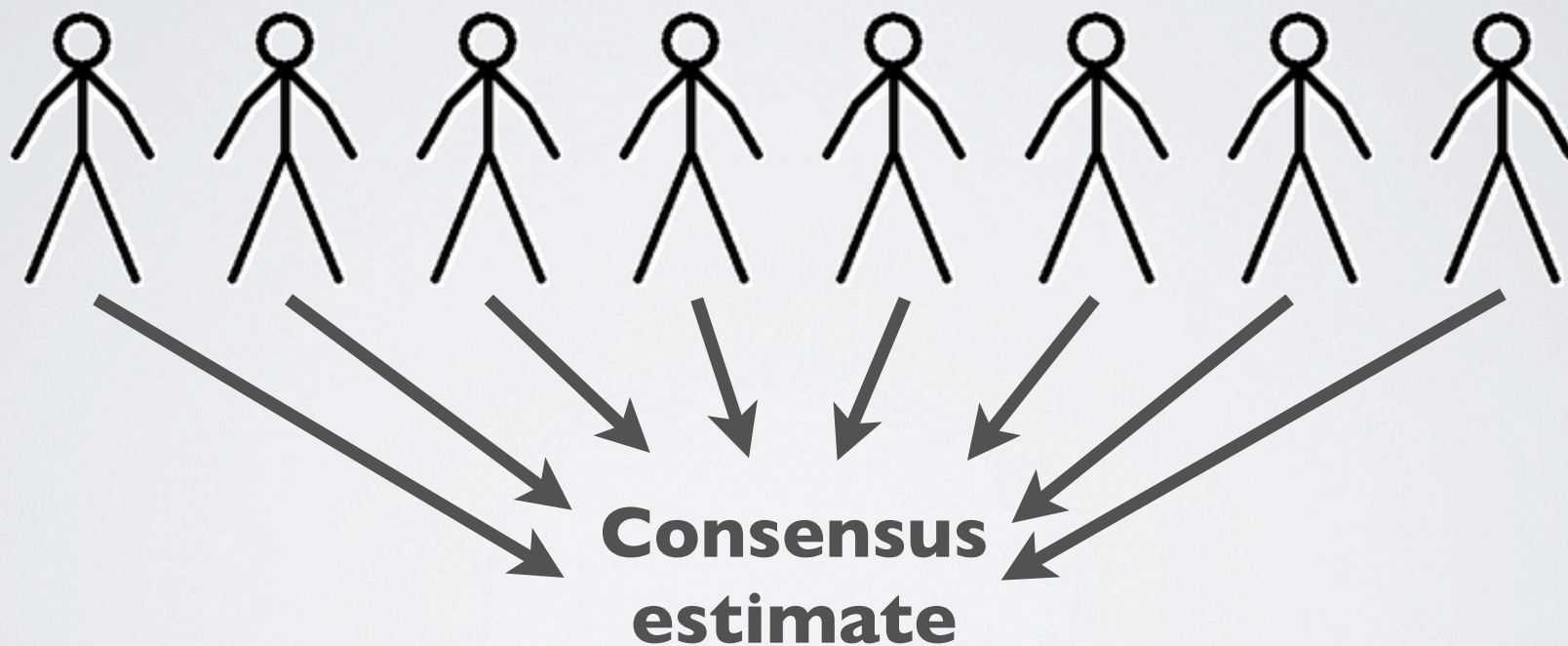
Joint work with Yiling Chen & Jenn Wortmann Vaughan



MY COAUTHORS

# RECAP: PREDICTION MARKETS

## MERGE BELIEFS



# BELIEFS, DISTRIBUTIONS, PRICES



# “PROBABILITY DOES NOT EXIST”

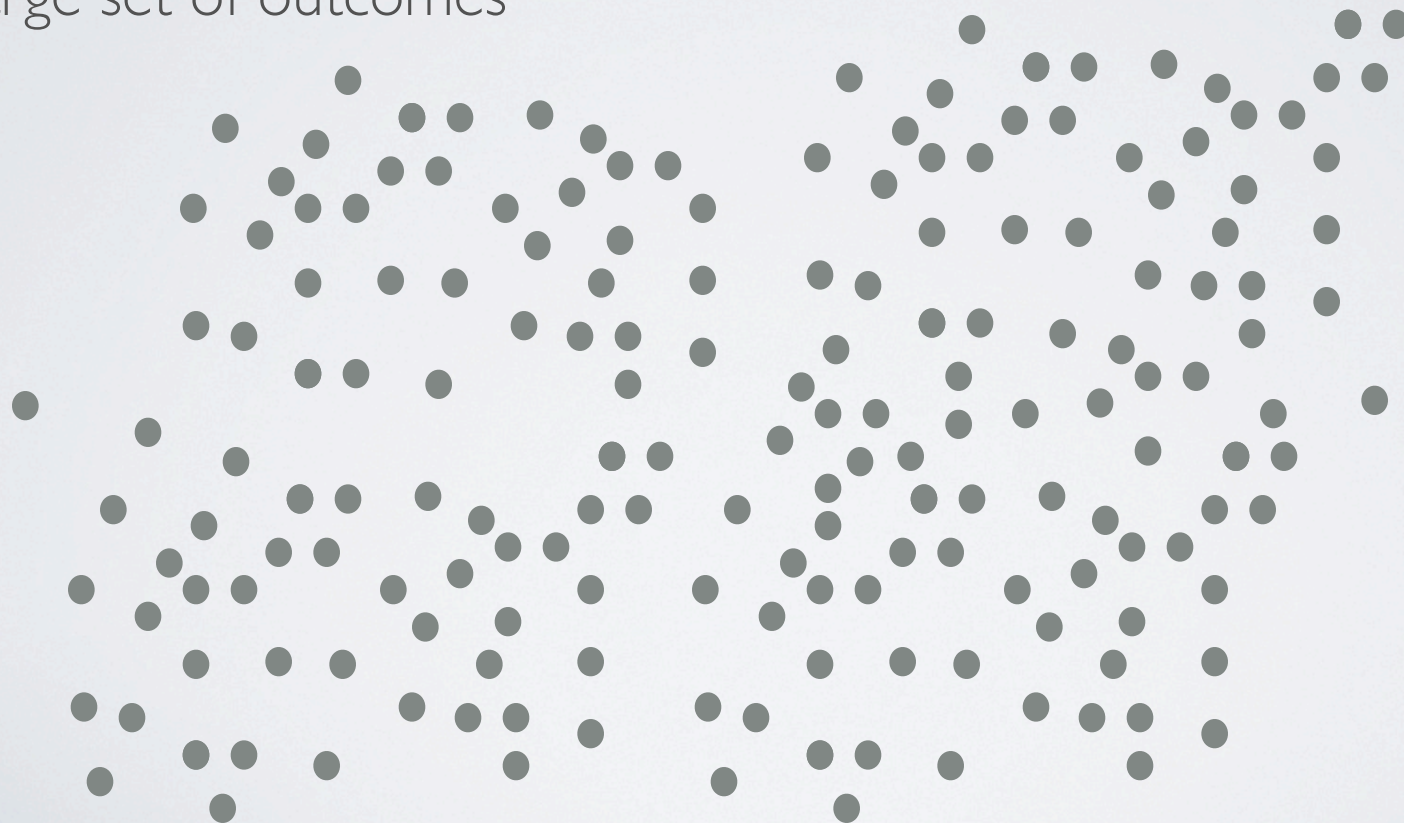
- The above phrase is what **Bruno de Finetti** wanted “printed in capital letters in the preface” to his Theory of Probability
- de Finetti: a probability  $P$  should be interpreted as the odds of a bet one would offer when our opponent can take any side of this bet.
- The laws of probability can be derived via simple “no-arbitrage conditions” of these odds

# TWO PROBLEMS FOR AUTOMATED MARKET MAKERS

- As contracts are purchased, how shall we set prices?
- How to handle *combinatorial* outcome spaces, i.e. when  $N$  is large?
  - Tournament outcome:  $N = n!$
  - Multi-candidate election:  $N = \binom{n}{k}$

# NAIVE APPROACH: ONE CONTRACT PER OUTCOME

- A natural strategy would just be to sell one contract for each of the large set of outcomes



# BETTER APPROACH: A SMALL “MENU” OF CONTRACTS

- Consider a multi-candidate election, where outcome is a set of  $k$  winners from  $n$  candidates,
- Market maker sells  $n$  contracts, one for each  $i$ , of the form:

**[pays off \$1 when  $i$  is among  $k$  winners]**

- That is, we allow bets on only a subset of “relevant” dimensions



# THE PAYOFF MATRIX

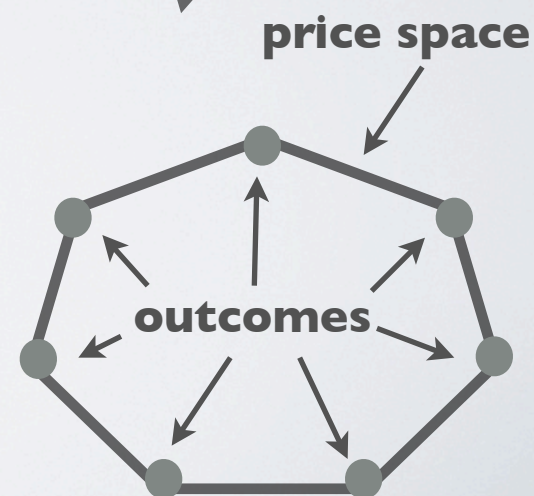
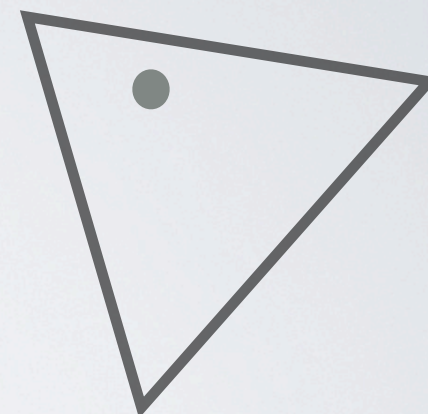
SMALL

L  
A  
R  
G  
E

	Cand. 1	Cand. 2	Cand. 3	Cand. 4	Cand. 5	Cand. 6
Outcome 1	\$1	\$0	\$1	\$1	\$0	\$0
Outcome 2	\$0	\$0	\$1	\$0	\$1	\$1
	.....					
Outcome N-1	\$0	\$0	\$1	\$1	\$1	\$0
Outcome N	\$0	\$0	\$0	\$1	\$1	\$1

# THE PRICE SPACE

- For simple markets, prices lie in the simplex:
- For “complex” markets, what constraints must we impose on the prices?
- Price vector must lie in  $\text{ConvexHull}(\text{outcomes})!$



# PRICING VIA REGULARIZATION: SIMPLE MARKETS

LMSR: 
$$\mathbf{p}(i) = \frac{\exp(\eta \mathbf{q}(i))}{\sum_j \exp(\eta \mathbf{q}(j))}$$

Alternative: 
$$\mathbf{p} := \arg \max_{\mathbf{p}' \in K} \mathbf{q} \cdot \mathbf{p}' - \frac{R(\mathbf{p}')}{\eta}$$

Neg. Entropy  $\downarrow$

Simplex  $\swarrow$

# PRICING VIA REGULARIZATION: COMPLEX MARKETS

$$\mathbf{p} := \arg \max_{\mathbf{p}' \in K} \mathbf{q} \cdot \mathbf{p}' - \frac{R(\mathbf{p}')}{\eta}$$

Some Curved  
Regularization  
↓

Convex Hull of  
Outcome Space  
↖

# RESULTS

- We have an efficient way to set prices in a prediction market with a combinatorial outcome space
- The “liquidity” (i.e. price stability) depends on the curvature properties of  $R$  -- more curved  $\Rightarrow$  more stability
- The worst-case loss of the market maker is no more than

$$\frac{\max_K R - \min_K R}{\eta}$$

# QUESTIONS?

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