Wages of Wins

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Motivation

- Previous work : predicting match outcomes
- Decent accuracies but what for ?
- MLSA '13 recommendation : beat the bookie !
- Could one actually make money ?

Setting

- Predictions : NCAAB « march madness », NBA, NFL – all 2015/2016
 - Outcomes, not scores Haven't managed yet to train good score predictors
 - Basketball as in earlier work, Football described later
- Data: www.kenpom.com, www.basketball-reference.com, www.pro-football-reference.com
- Money Lines : www.vegasinsider.com

Money Lines

- Goal for sports book : make money !
 - Partition betting volume s.t.
 bettors' wins (+profit) covered by losses
 - Adjust « money lines »

Money Lines (2)

	Match-Up	Favorite	Underdog	Fav-Line	DogLine
1	Detroit at Atlanta	Atlanta	Detroit	300	240
2	Utah at Detroit	Detroit	Utah	110	-110

- Match 1 :
 - Correct bet Detroit, 100\$ bet = 240\$ gain

Atlanta clear favorite : motivate people to bet on Detroit

- Correct bet Atlanta, 300\$ needed to win 100\$ → 100\$ bet = 33.33\$ gain
- Match 2 : « Pick 'em »
 - Correct bet of 100\$ = 90.90\$ gain
- Initially based on model, bettor behavior shifts lines

Correct Dog-Bet > Correct Pick 'em-Bet > Correct Fav-Bet

Remember : wins + PROFIT

Simulated betting

- Bet all matches
 - 100\$ per match
 - Separate regular/post-season for NBA/NFL
- At money-line w/smallest spread
 - E.g. 170/150 instead of 200/170

Conservative but NOT as conservative as I thought !

- Tally winnings per day
 - And over full period

« Vegas accuracy »

- « Predicts » only favorites
- Pick 'ems → Coin flips →
 50 % correct expected
 - Best case all correct

Can make big difference financially

- Worst case all wrong

NCAAB (67 matches)

5 Pick 'ems

<u>Vegas</u> w/ Pick 'ems w/o Pick 'ems Pay-out Accuracy Pay-out Exp. Acc Best Acc. Pay-out Worst Pay-out 0.7419 30.26 0.7313 7.51 0.7611 484.76 0.6865 -469.7346 Favs **Predictors Naïve Bayes Multilayer Perceptron Simplified KenPom** Classifier Accuracy 0.6865 0.6417 0.7014 Pay-out 293.52 -605.92-231.34

Distribution of predictions

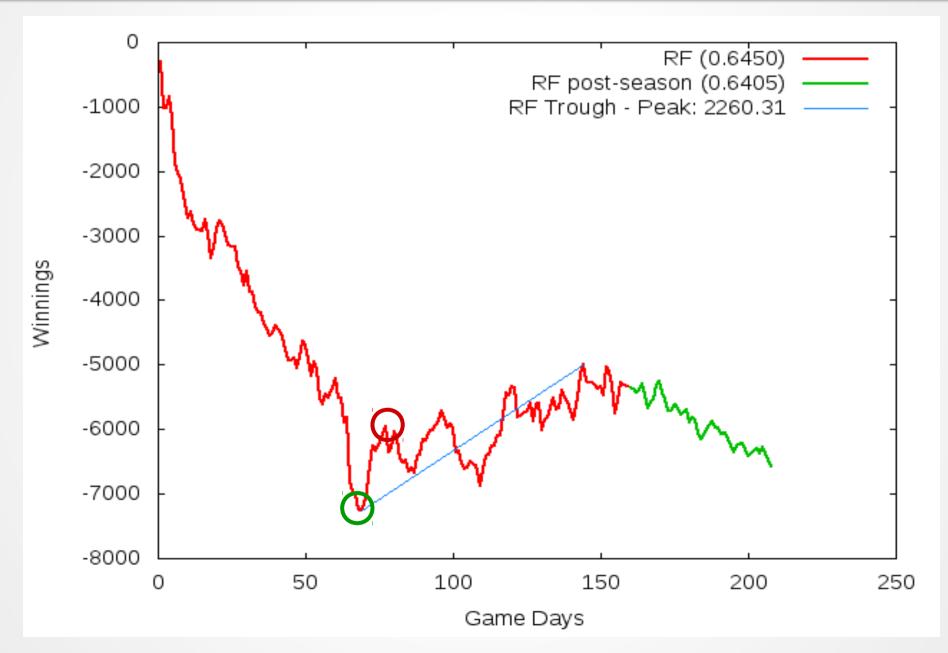
Classifier	Favs	Dogs	PEs
NB	39	5	2
MLP	38	2	3
KP	43	0	4

NBA (1288, 115 PEs)

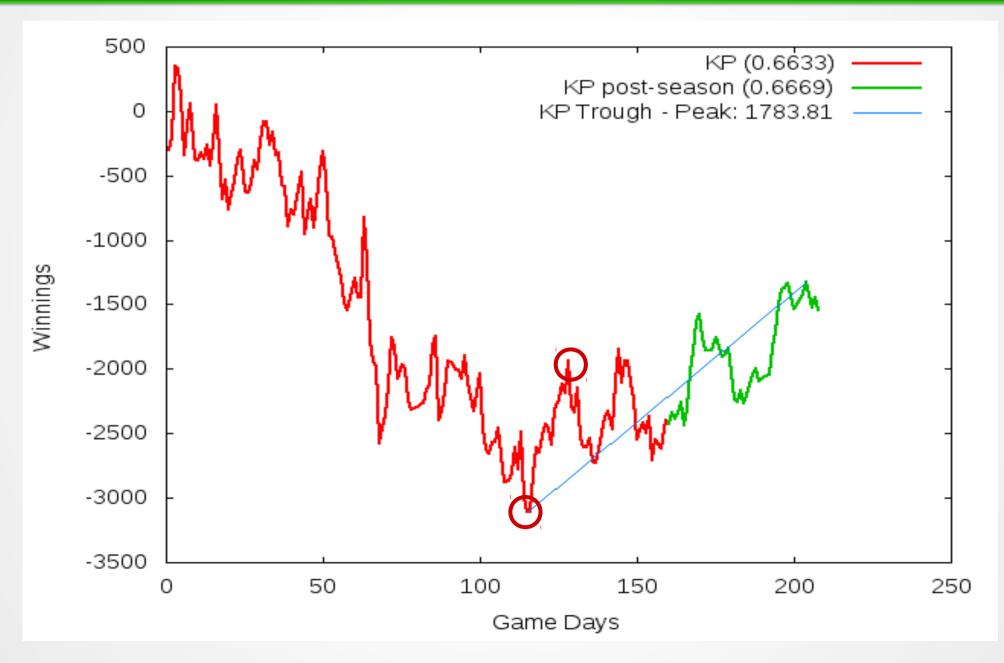
<u>Vegas</u>

w/	o Pick 'em	IS				w/	Pic	k 'ems				
Accuracy Pay-out		out Ex	кр. Асс	Pay-out		Best Acc.		Pay-out	Worst.		Pay-out	
0.7122	0.7121 -2374.1		6865	-1857.3		0.7375	1	9125.84	0.6492		-12828.81	
835 Favs			<u>Predictors</u>									
	Classifier		NB			MLP	Ra	Random Forest		KP		
	Accuracy		607		0.6	615	0.6405			0.6669		
	Pay-out		82.21			05.56	66 -6580.88		-1543.05		3.05	
		-	<u>Distri</u>	butior	<u>1 0</u>	of pre	dic	<u>tions</u>		jular + t-seas		
			ssifier	Favs		Dog	S	PEs				
		NB		740		62		49				
			C	764		66		22				
		RF		732		65		28				
		KP		783		64		12				

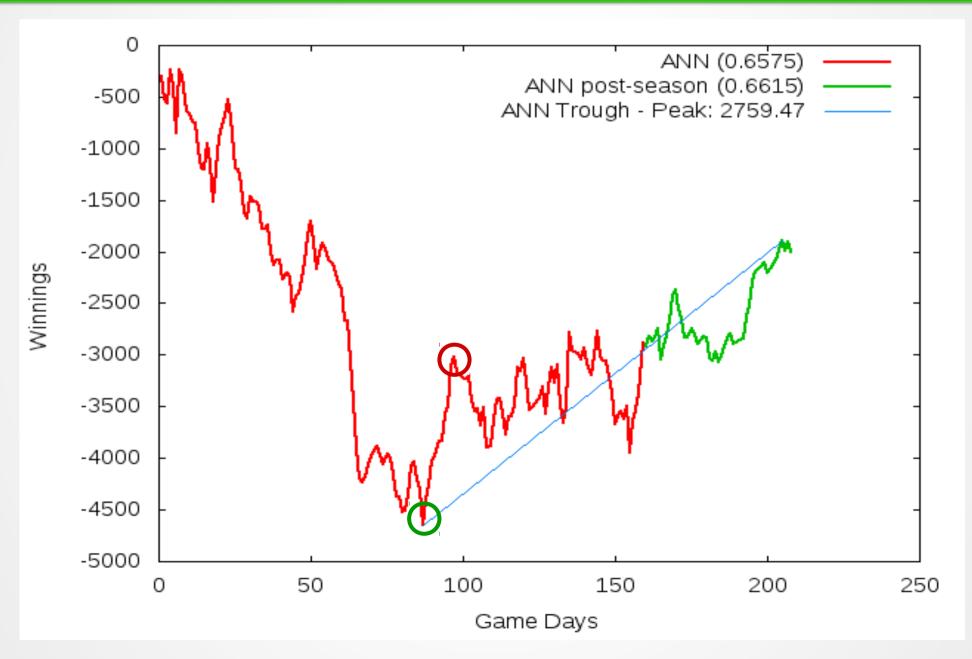
RF winnings curve



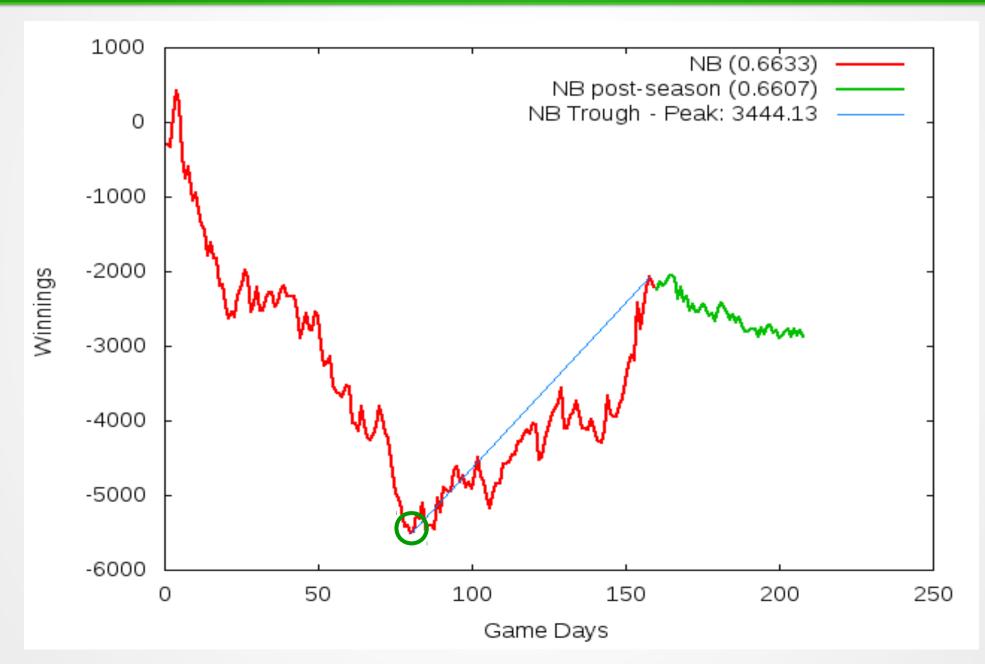
KP winning curve



MLP winning curve



NB winning curve



NFL prediction

- Different game :
 - Different outcomes for possessions
 - Not many scoring events
 - Two (three) distinct sub-teams
 - Fewer matches
- State of the art : « Defense-adjusted Yards above Replacement » (DYAR)
 - Needs fine-grained data, expensive calculation
- My « solutions » :
 - Averaged (adjusted) statistics
 - Opponents' statistics
 - Strength of schedule

NFL (251, 29 PEs)

<u>Vegas</u>

	w/o Pic	ck 'ems		w/ Pick 'ems						
	Accuracy	Pay-out	Exp. Acc	c Pay-out	Best Acc.	Pay-out	Worst	Pay-out		
	0.6441	-1215.69	0.6294	-1251.92	0.6852	1420.68	0.5697	-4115.42		
143 Favs Predictors										
	Classifi	ier	NB	MLP	RF	Sim	ple Rating	y System		
Accuracy			0.6335	0.5896	0.5737	0.5896				
Pa	iy-out		1777.98	729.92	-1591.68	-1255.71				

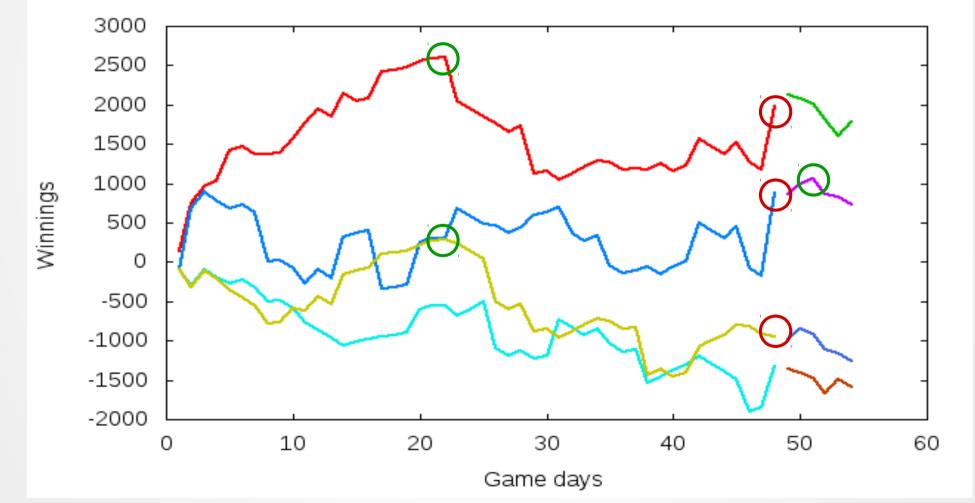
Distribution of predictions

Classifier	Favs	Dogs	PEs
NB	119	26	14
MLP	103	29	16
RF	111	17	16
SRS	115	18	15

NFL winning curves

- RF (0.5792) -----
- RF post-season (0.5737) ----
 - SRS (0.5958) -----
- SRS post-season (0.5896)

- NB (0.6417)
- NB post-season (0.6335)
 - ANN (0.5917) -----
- ANN post-season (0.5896) -



Can an amateur make money ?

- « Jein »
- NCAAB : possible but few matches/upsets → risky
- NBA : know when to get in
- NFL : know when to get out

Use Naïve Bayes !

Future work

Not all

Favs/Dogs

are equal !

 Characterize distribution by confidence

Optimize gains instead of accuracy

- How much to bet
- Learn strategies for which
 matches to bet on
 Exists for
 soccer

I assume no liability for betting losses !

Work-in-progress, unpublished things (like NFL methodology) :

http://scientificdm.wordpress.com