

Piece rates, salaries and tournaments: economic and psychological competition in a real effort task

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How should you pay someone?

- Piece rates
 - Output – Pay based on absolute performance
 - Input – Pay based on time
- Salary
 - Pay independent of performance; no **extrinsic** incentive
- Tournaments
 - Pay based on relative performance;
 - Winner take all; Rank-order
 - *Lazear and Rosen (1981); Green and Stokey, (1983); Nalebuff and Stiglitz (1983)*

Voluminous literature looking at incentive mechanisms

- Field studies
 - Executive positions (Gibbons and Murphy, 1990), Chicken farmers (Knoeber, 1989, 1994), Law firms (Ferrall, 1996), Portfolio managers (Brown, 1996), Executives (Xu, 1997), Etc.
- Experimental work
 - Bull, Schotter and Weigelt (1987)
 - Harbring and Irlenbusch (2003)
 - Van Dijk et al (2001)
- Surveys
 - Prendergast (1999), Dechenaux, Kovenock, Sheremeta (2012)

Three ways tournaments differ from piece rates

- ***Psychological competition***
 - Act of competing against another 
 - pleasure of winning/ pain of losing
- ***Economic competition***
 - Performance dependent payoffs
 - “Bad” decisions leading to low payoff
- ***Information***
 - Information about how well you are doing
 - How well it is possible to do
 - Provides a reference for identifying attainable level

Objectives of our study

- ***Study performance under different incentive schemes***
- Using a “real effort task”
- Cognitively difficult task ->
- People need to expend “effort” to perform the task well

The task used in this study

- Multiple Cue Probabilistic Learning Task
- In each round subjects asked to forecast price of fictitious “stock” given two cue-values A and B
- Stock price:
- $P_t^* = 10 + 0.3 * CUE A_t + 0.7 * CUE B_t + e_t$

The task used in this study

- Cue values change each round, but **not** the underlying relationship
- ***Metric of good decision (good performance) ->***
- ***Absolute forecasting error:***
- **$e_{it} = |P_t^* - P_{it}|$**

Experimental design

- Computerized experiments
- \$5 show up fee
- Instructions read out loud
- 5 minutes to study 10 examples of Price/Cue relationship provided on paper
- Experiment starts after that

Experimental design

- Shown Cue A and Cue B for 1st round
- Given time to enter decisions
- Results displayed
- New cue values for second round and so on
- Continue for 20 rounds

Three types of payment schemes

- **Piece rate:**
- $Earning_{it} = \$1.00 - e_{it}$

- **Two person (winner take all) tournament:**
- $Earning_{it} = \$1.00$ if $|e_{it}| < |e_{jt}|$
 $= \$0.00$ otherwise

- **Salary**
- Earnings = \$20 (announced before-hand)

Treatments

- ***Piece rates***
- ***Piece rates with win-loss information (Win-Lose)***
- ***Tournament***
- ***Tournament no information***
- ***Salary***
- ***Salary with win-loss information***

Piece rates

- Rounds 1 – 20
- Payment based on own absolute errors only
- $Earning_{it} = \$1.00 - e_{it}$
- If absolute error > 100 then receive \$0

Win-Lose

- **Rounds 1 – 5**
 - Piece rate payment scheme exactly as before

 - **Rounds 6 – 20**
 - Assigned partner each round; **anonymous**
 - Partners randomly re-matched each period,
 - Same piece rate payment scheme but
 - At the end of the round subjects learn
- 1. Earnings**
 - 2. WIN or LOSE** (*whether one's own error was smaller (larger) than pair member's error*)

Tournament

- ***Rounds 1 – 5***
- Piece rate payment scheme exactly as before

- ***Rounds 6 - 20***
- Assigned partner each round
- Provided extra \$4.00 in earnings account
- At the end of the round subjects learn
 - 1. Error***
 - 2. WIN or LOSE***
 - 3. Payment = \$1.00 or \$0.00***

Tournament no information

- ***Rounds 1 – 5***
 - Piece rate payment scheme exactly as before

 - ***Rounds 6 - 20***
 - Assigned partner each round
 - Provided extra \$4.00 in earnings account
 - At the end of the round subjects learn
- 1. Error***
- At the end of ROUND 20 subjects learn
- 1. WIN or LOSE***
- 2. Payment = \$1.00 or \$0.00 for each round***

Salary

- Rounds 1 - 20
- Flat \$20 payment announced at the beginning
- Shown earnings based on Piece Rate
- At the end of the round subjects learn

1. Error

2. Earnings

- But made clear that they receive a flat amount at the end regardless of errors or “per round earnings”

Salary with Win-Loss

- Rounds 1 - 20
- Flat \$20 payment announced at the beginning
- Shown earnings based on Piece Rate
- At the end of the round subjects learn
- **Error**
- **Earnings**
- **Win or Lose**
- But made clear that they receive a flat amount at the end regardless of errors or “per round earnings”

Questions

1. Is winning/losing important in pay for performance schemes?

- Piece rate vs. Win/Lose
- Incentives the same, information different

2. Are payoffs important when paying for performance?

- Win/Lose vs. Tournament
- Information same, incentives different

Questions

3. Is information in tournaments important?

- Tournament vs. Tournament No Info
- Incentives the same, but info different

4. Extrinsic versus intrinsic motivation

- Compare piece rate with salary

5. Is winning/losing important when pay is independent of performance?

- Compare Salary vs. Salary Win-Lose

Questions

- Collect demographic information along with gender
- Prior to start of game, we measure
 - 1. Trait Anxiety**
- Following game, we measure
 - 1. Motivation**
 - 2. Effort**
 - 3. Competence**
 - 4. Interest**

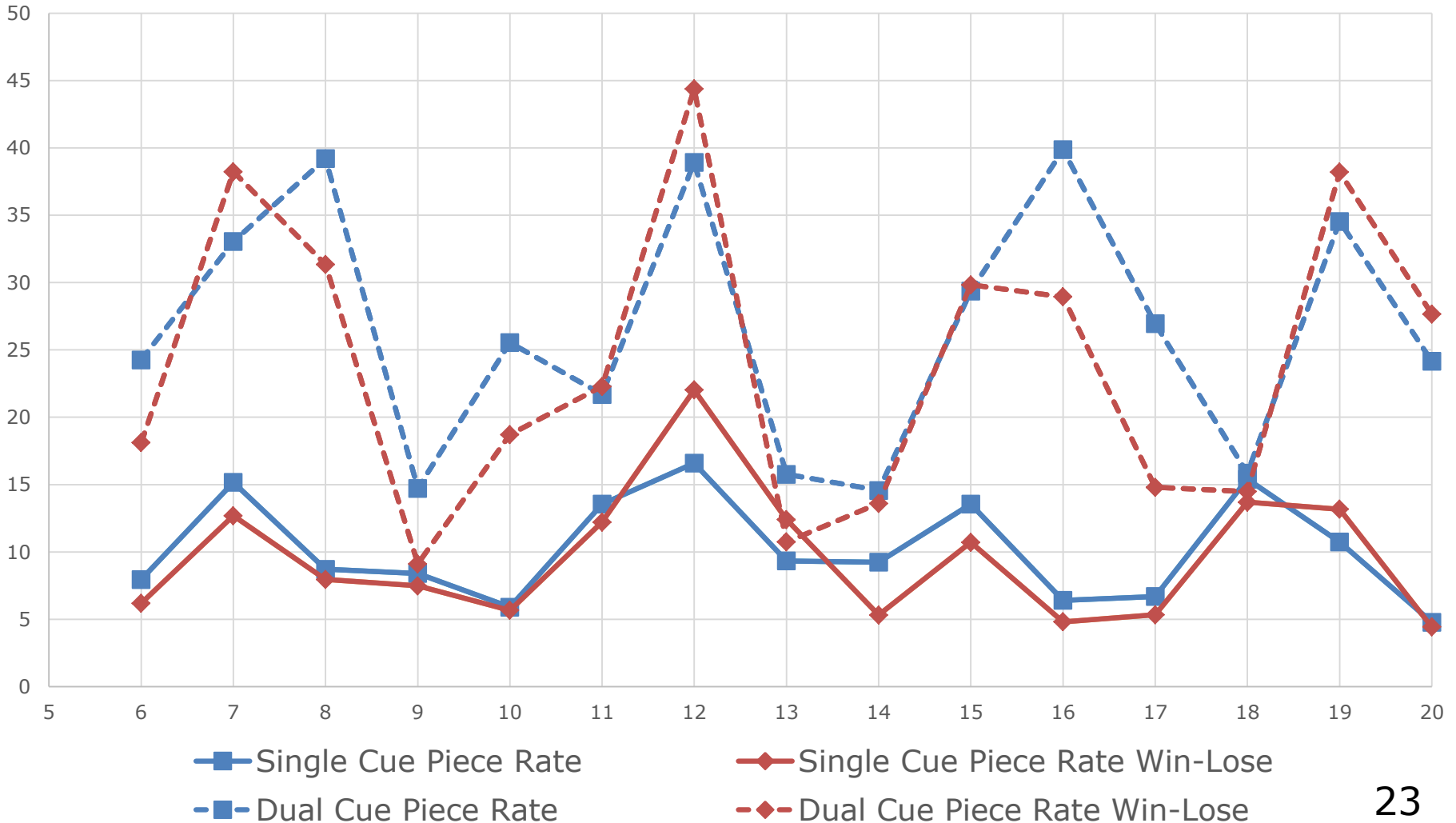
Experimental design

- Two separate experiments with 376 subjects
- *Experiment #1 with 176 subjects*
- **Here both cue values change from one round to the next**
- *Experiment #2 with 200 subjects*
- Here Cue A **fixed at 150**; only Cue B changes

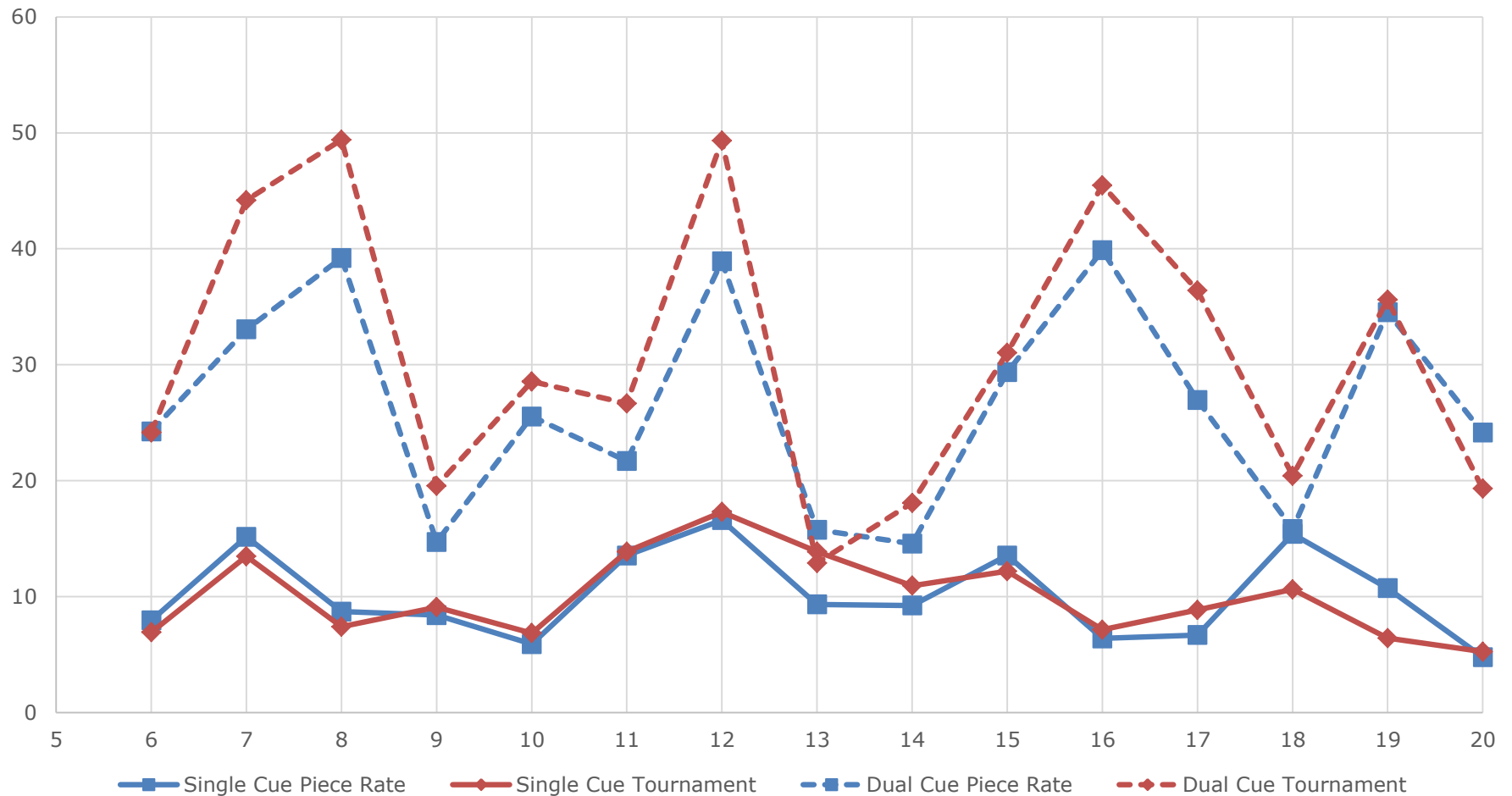
Overview of Results

	<i>Single Cue</i>		<i>Dual Cue</i>		<i>Overall</i>
	n	Average Errors	n	Average Errors	Average Errors
<i>Piece Rate</i>	42	10.2	39	26.6	18.1
<i>Win-Loss</i>	42	9.6	35	24.0	16.2
<i>Tournament</i>	40	10.0	38	30.7	20.1
<i>Salary</i>	42	9.0	34	25.1	16.2
<i>Salary Win-Loss</i>	34	10.2	30	31.4	20.2

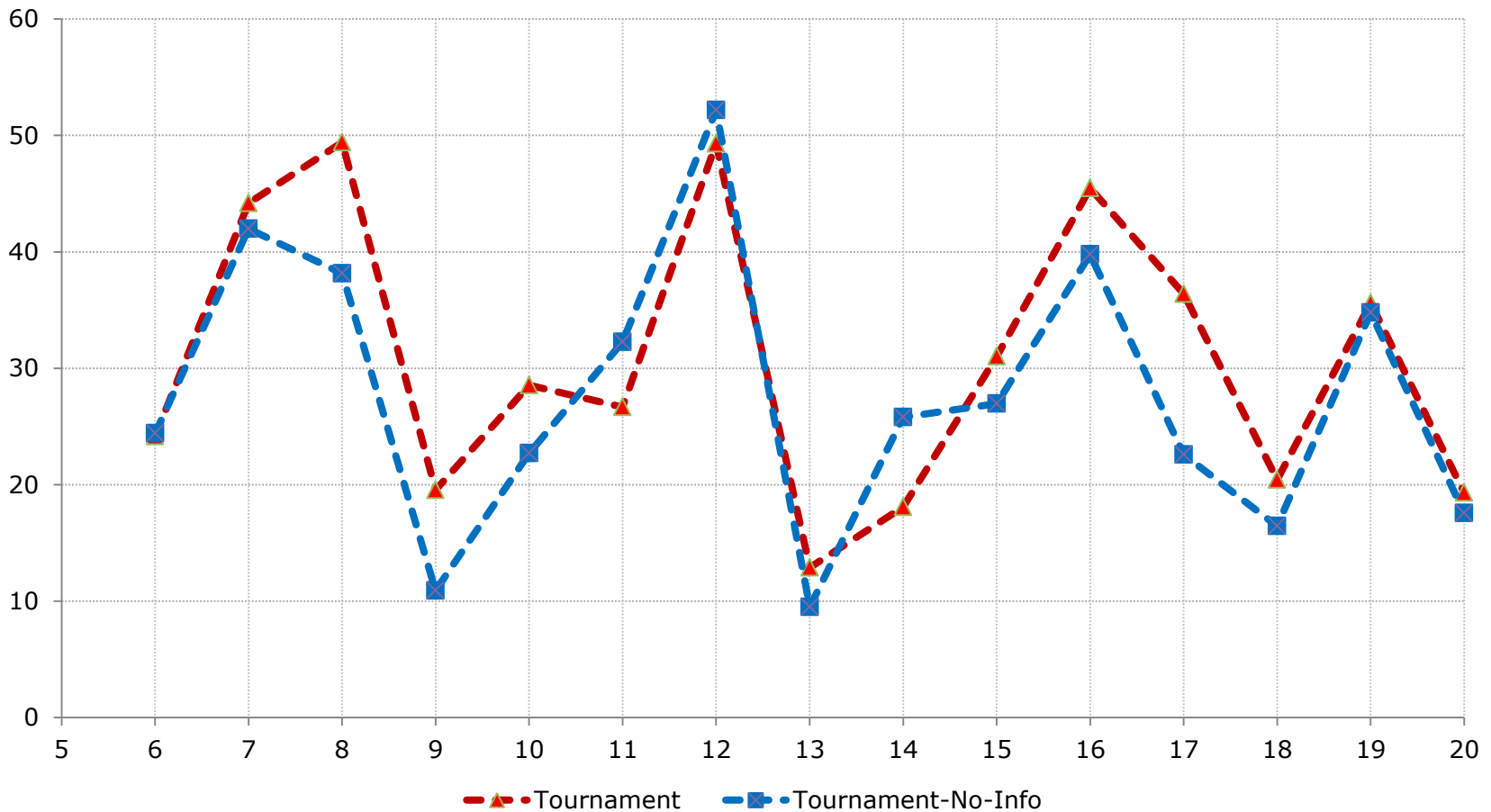
Piece Rate vs Piece-Rate Win-Lose



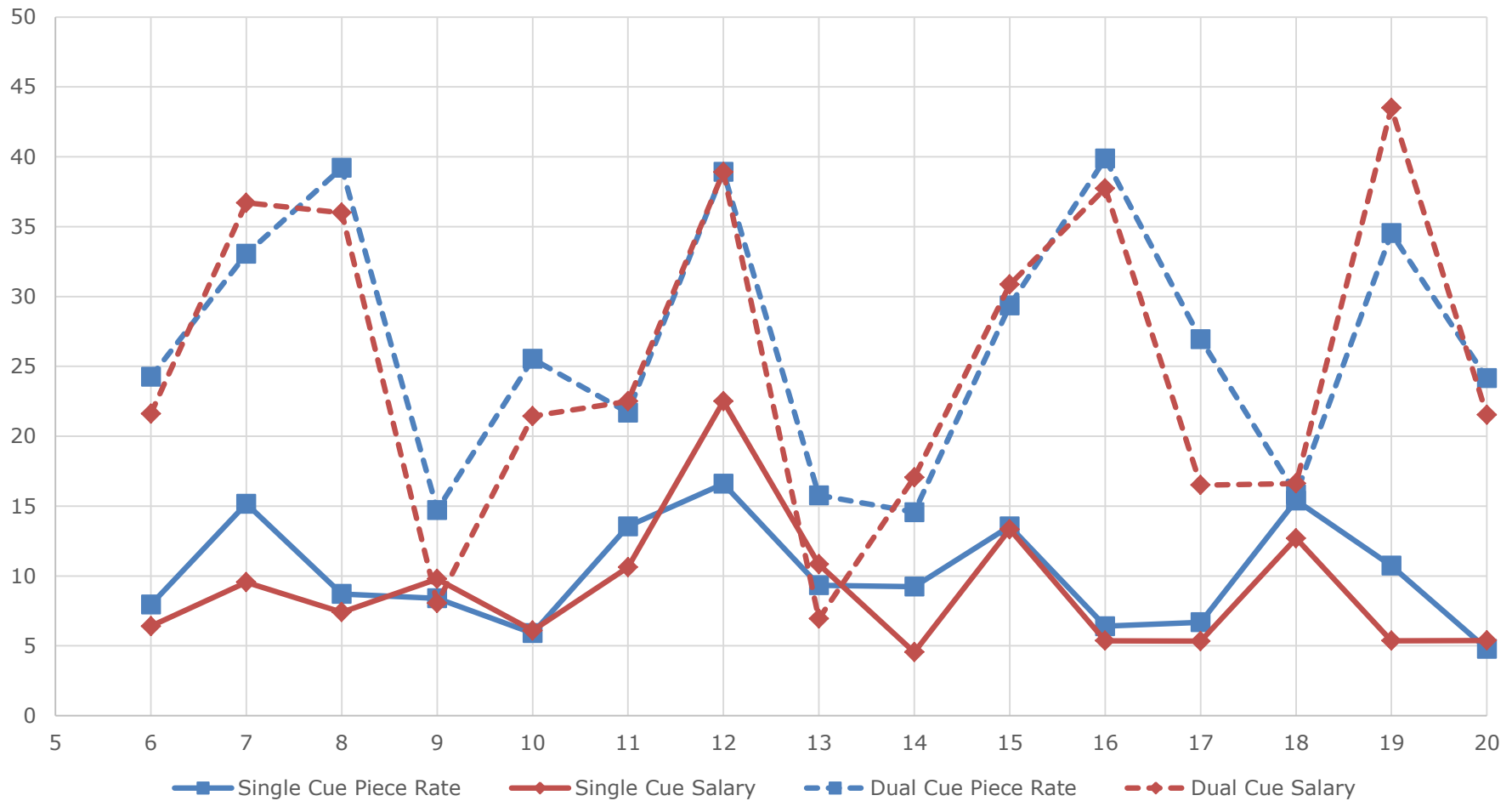
Piece-Rate Win-Lose vs Tournament



Dual Cue Tournament Vs Tournament No Information



Piece Rate vs Salary



Salary vs Salary-Win-Lose



Dependent variable: Absolute forecast error (All Data)

	Model 1	Model 2	Model 3	Model 4
Piece Rate Win-Lose	-1.84 (2.10)	-1.79 (2.03)	-3.92* (2.23)	-5.49* (2.99)
Tournament	2.04 (2.66)	1.15 (2.68)	0.10 (2.90)	3.14 (3.16)
Salary	-1.80 (2.09)	-4.21 (2.78)	-6.00** (2.91)	-7.34** (3.00)
Salary Win-Lose	2.13 (2.62)	-0.29 (3.20)	-1.55 (3.46)	-2.84 (3.30)
Cuespread	0.08*** (0.00)	0.08*** (0.01)	0.08*** (0.01)	0.08*** (0.01)
Lagged Earnings		-2.94 * (1.60)	-2.50 (1.64)	-2.63 (1.65)
Trait Anxiety			0.10 (0.11)	0.10 (0.11)
Female			6.46*** (1.35)	6.45*** (1.34)
Round	0.17 *** (0.06)	0.17 *** (0.06)	0.17 ** (0.07)	0.15 (0.14)
Constant	6.39 *** (1.91)	8.48 *** (2.66)	2.01 (4.34)	2.27 (4.61)
With treatment-round interactions	No	No	No	Yes
Observations	5640	5640	5130	5130
Participants	376	376	342	342
R²	0.082	0.090	0.110	0.110

p-values for Wald χ^2 test on treatment dummy coefficients

	Model 1	Model 2	Model 3	Model 4
P(PRWL = 0)	0.38	0.379	0.079	0.067
P(PRWL = T)	0.093	0.184	0.076	0.008
P(PRWL = S)	0.977	0.252	0.310	0.552
P (S = 0)	0.391	0.13	0.04	0.015
P (S = SWL)	0.082	0.082	0.06	0.09

Dependent Variable: Absolute Forecast error (Dual cue)

	Model 1	Model 2	Model 3	Model 4
Piece Rate Win-Lose	-2.53 (3.43)	-2.43 (3.28)	-4.26 (3.35)	-6.51 (4.82)
Tournament	4.17 (4.41)	3.00 (4.35)	1.68 (4.57)	6.86 (4.79)
Salary	-1.50 (3.40)	-5.27 (4.40)	-7.28 (4.54)	-10.22 ** (4.95)
Salary Win-Lose	4.87 (4.32)	1.10 (5.14)	2.07 (5.37)	-4.42 (5.03)
Cuespread	0.11 *** (0.01)	0.11 *** (0.01)	0.12 *** (0.01)	0.12 *** (0.01)
Lagged Earnings		-5.06 * (2.65)	-3.80 (2.57)	-3.95 (2.61)
Trait Anxiety			0.07 (0.17)	0.07 (0.17)
Female			8.69 *** (2.14)	8.68 *** (2.14)
Round	0.52 *** (0.11)	0.52 *** (0.11)	0.54 *** (0.12)	0.47 * (0.24)
Constant	3.71 (3.30)	7.43 * (4.39)	-0.85 (6.85)	0.16 (7.26)
With treatment-round interactions	No	No	No	Yes
Observations	2640	2640	2430	2430
Participants	176	176	162	162
R²	0.107	0.110	0.137	0.138
p(PRWL = T)	0.069	0.119	0.092	0.013
p(S = 0)	0.660	0.231	0.109	0.039
p(S = SWL)	0.073	0.073	0.017	0.249

Dependent Variable: Absolute Forecast error (Single cue)

Dep Var: Forecast Error	Single Cue	Single Cue	Single Cue	Single Cue
Piece Rate Win-Lose	-0.56 (1.49)	-0.55 (1.48)	-1.32 (1.71)	-2.88 (3.31)
Tournament	-0.15 (1.66)	-0.56 (1.64)	-0.85 (1.95)	-0.15 (3.25)
Salary	-1.15 (1.35)	-2.16 (1.71)	-3.29* (1.88)	-3.84 (3.08)
Salary Win-Lose	0.06 (1.47)	-0.95 (1.81)	-2.29 (1.95)	0.47 (3.67)
Cuespread	0.02 *** (0.00)	0.02 *** (0.00)	0.02 *** (0.00)	0.02 *** (0.00)
Lagged Earnings		-1.13 (1.14)	-1.41 (1.25)	-1.44 (1.24)
Trait Anxiety			0.03 (0.09)	0.03 (0.09)
Female			1.31 (1.01)	1.30 (1.01)
Round	-0.10 * (0.06)	-0.10 * (0.06)	-0.12 ** (0.06)	-0.11 (0.14)
Constant	8.74 *** (1.46)	9.79 *** (1.78)	9.35 ** (4.19)	9.22 * (4.87)
With treatment-round interactions	No	No	No	Yes
Observations	3000	3000	2700	2700
Participants	200	200	180	180
R²	0.018	0.020	0.025	0.026
p(S = 0)	0.394	0.206	0.080	0.211

High versus low performers

Combined data

- Why do tournaments not perform well in general?
- It appears that those who are good at the task perform about the same in all treatments
- But those who are not good perform better in Win-Lose and Salary
- What does it mean to say – good or bad at the task?

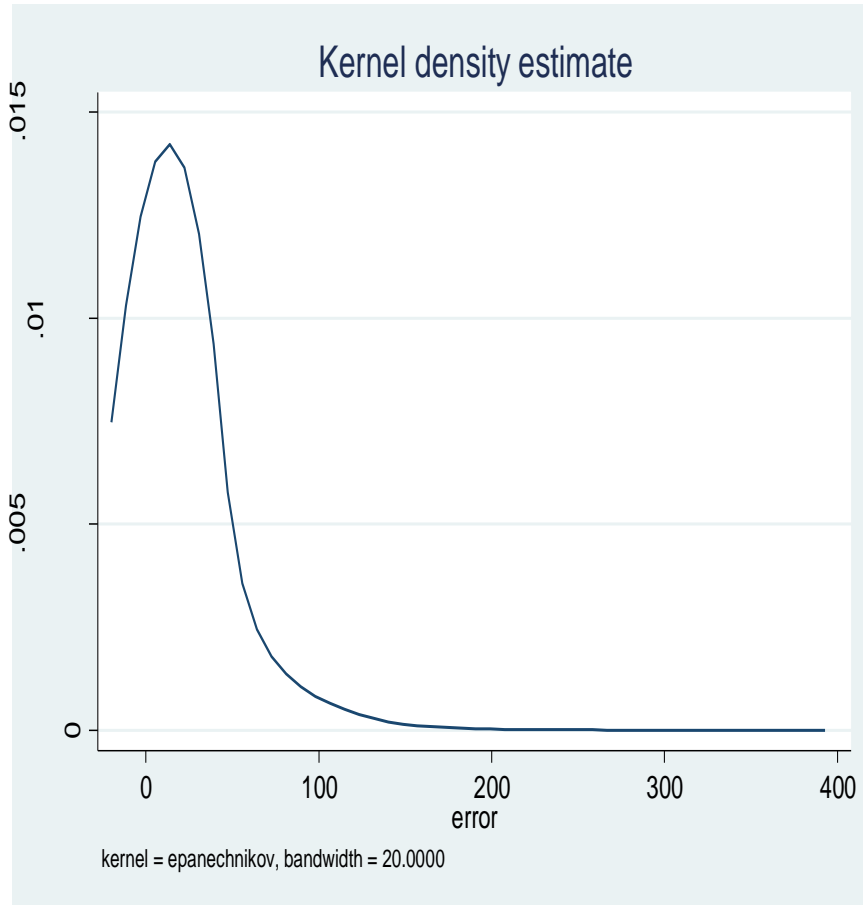
High versus low performers

Combined data

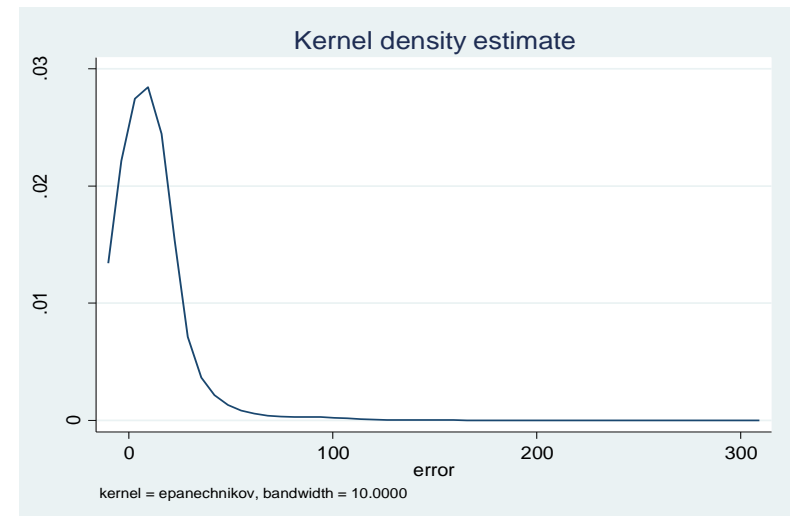
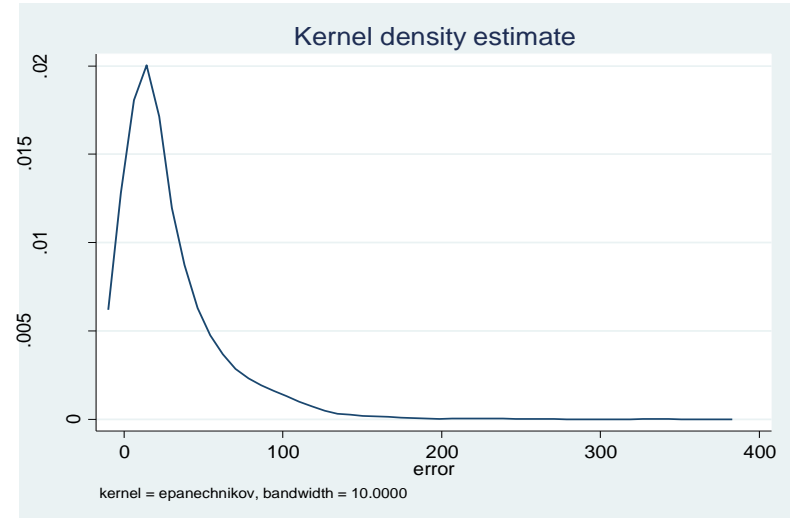
- Because everyone plays under a piece-rate condition during the first five rounds, we can look at performance in those rounds to split people up into “high” and “low” performers
- Split by Mean or Median?
 - *Data positive/right skewed (long right tail)*
- Below we present results for combined data and those above and below the mean
 - ***Mean (Median) error or higher -> LOW performer***
 - ***Lower than Mean (Median) -> HIGH performer***

Kernel Density Plots

DUAL CUE



ALL



SINGLE CUE

Regression of errors for high performers (mean split)

Dep Var: Forecast Error	High Perf	High Perf	High Perf	High Perf
Piece Rate Win-Lose	2.63 *	2.55 *	1.50	2.19
	(1.48)	(1.45)	(1.59)	(2.43)
Tournament	2.14	1.30	1.23	3.73
	(1.68)	(1.59)	(1.76)	(2.65)
Salary	0.76	0.74	0.09	0.80
	(1.35)	(1.31)	(1.36)	(2.18)
Salary Win-Lose	-0.17	-0.15	-1.31	1.07
	(1.26)	(1.23)	(1.34)	(2.57)
Cuespread	0.05 ***	0.05 ***	0.05 ***	0.05 ***
	(0.00)	(0.01)	(0.01)	(0.01)
Lagged Earnings		-2.54 *	-2.58 *	-2.65 *
		(1.43)	(1.55)	(1.56)
Trait Anxiety			0.08	0.08
			(0.07)	(0.07)
Female			2.71 **	2.71 **
			(1.06)	(1.06)
Round	-0.03	-0.03	-0.06	0.03
	(0.05)	(0.05)	(0.05)	(0.09)
Constant	4.61 ***	6.68 ***	3.07	1.95
	(1.13)	(1.74)	(3.37)	(3.63)
With treatment-round interactions	No	No	No	Yes
Observations	3240	3240	2940	2940
Participants	216	216	196	196
R²	0.079	0.085	0.100	0.101
p(PRWL = 0)	0.076	0.078	0.346	0.367
p(PRWL = T)	0.787	0.452	0.873	0.560
p(PRWL = SWL)	0.044	0.045	0.029	0.662
p(S = 0)	0.572	0.570	0.949	0.716
p(S = SWL)	0.456	0.460	0.192	0.906

Regression of errors for low performers (mean split)

Dep Var: Forecast Error	Low Perf	Low Perf	Low Perf	Low Perf
Piece Rate Win-Lose	-4.54 (3.77)	-3.99 (3.37)	-6.24 * (3.60)	-12.45 * (6.38)
Tournament	3.32 (4.81)	0.18 (4.54)	-1.74 (4.72)	2.57 (5.30)
Salary	-4.26 (3.55)	-3.83 (3.17)	-7.17 ** (3.62)	-10.71 ** (4.75)
Salary Win-Lose	0.63 (4.19)	0.58 (3.82)	0.24 (4.12)	-3.53 (4.50)
Cuespread	0.11 *** (0.01)	0.11 *** (0.01)	0.11 *** (0.01)	0.11 *** (0.01)
Lagged Earnings		-11.36 *** (3.44)	-10.63 *** (3.39)	-10.61 *** (3.42)
Trait Anxiety			0.06 (0.15)	0.06 (0.15)
Female			5.91 *** (2.31)	5.92 ** (2.31)
Round	0.45 *** (0.13)	0.45 *** (0.12)	0.48 *** (0.13)	0.36 (0.29)
Constant	8.23 ** (3.53)	15.95 *** (4.83)	9.96 (6.35)	11.53 * (6.90)
With treatment-round interactions	No	No	No	Yes
Observations	2400	2400	2190	2190
Participants	160	160	146	146
R²	0.091	0.109	0.121	0.122
p(PRWL = 0)	0.229	0.237	0.082	0.051
p(PRWL = T)	0.066	0.270	0.244	0.032
p(PRWL = SWL)	0.147	0.155	0.052	0.154
p(S = 0)	0.230	0.228	0.048	0.024
p(S = SWL)	0.142	0.141	0.027	0.121

Regression of errors for high performers (median split)

Dep Var: Forecast Error	High Perf	High Perf	High Perf	High Perf
Piece Rate Win-Lose	1.58 (1.09)	1.58 (1.09)	1.16 (1.26)	2.15 (1.78)
Tournament	0.96 (1.22)	1.01 (1.26)	1.12 (1.55)	1.14 (2.46)
Salary	1.98 * (1.16)	1.98 * (1.15)	1.24 (1.07)	2.14 (2.01)
Salary Win-Lose	-0.04 (0.74)	-0.04 (0.74)	-0.39 (0.93)	-0.51 (1.70)
Cuespread	0.03 *** (0.00)	0.03 *** (0.00)	0.03 *** (0.00)	0.03 *** (0.00)
Lagged Earnings		0.15 (1.26)	0.44 (1.44)	0.47 (1.43)
Trait Anxiety			0.03 (0.05)	0.03 (0.05)
Female			1.22 (0.95)	1.22 (0.96)
Round	0.00 (0.04)	0.00 (0.04)	-0.03 (0.04)	0.01 (0.09)
Constant	4.11 *** (0.77)	3.87 *** (1.29)	2.27 (2.32)	1.80 (2.70)
With treatment-round interactions	No	No	No	Yes
Observations	1950	1950	1740	1740
Participants	130	130	116	116
R²	0.048	0.048	0.053	0.053
p(PRWL = 0)	0.148	0.148	0.358	0.229
p(PRWL = T)	0.672	0.706	0.981	0.676
p(PRWL = SWL)	0.138	0.138	0.190	0.105
p(S = 0)	0.086	0.086	0.249	0.289
p(S = SWL)	0.080	0.080	0.125	0.169

Regression of errors for low performers (median split)

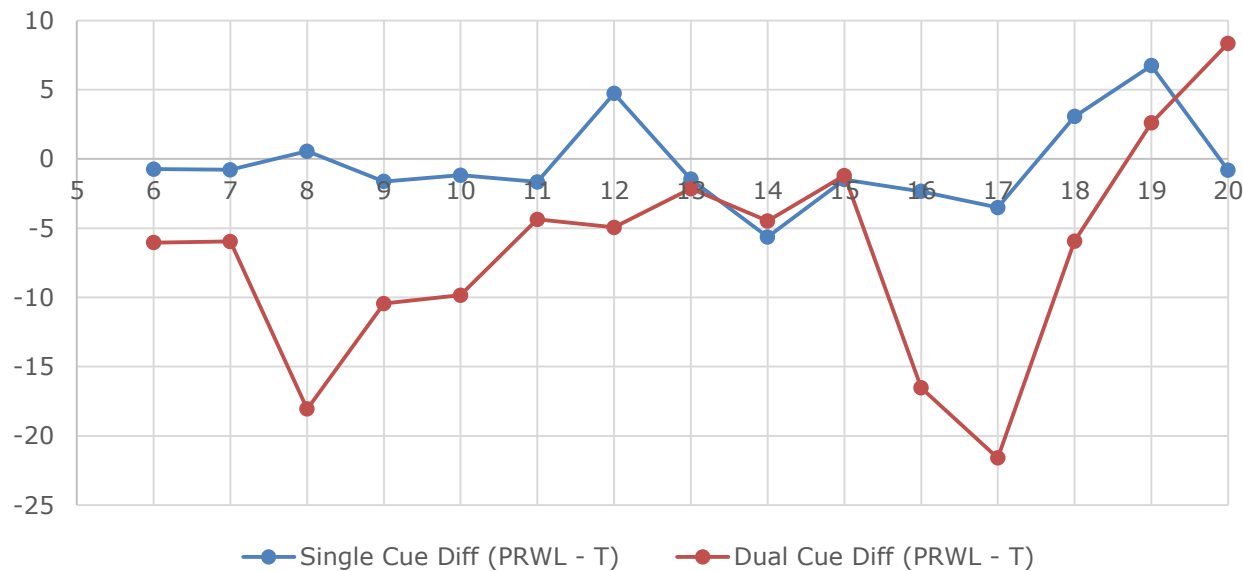
Dep Var: Forecast Error	Low Perf	Low Perf	Low Perf	Low Perf
Piece Rate Win-Lose	-3.49 (2.78)	-3.15 (2.53)	-4.78* (2.66)	-7.43* (4.05)
Tournament	2.30 (3.52)	-0.34 (3.38)	-1.50 (3.50)	3.27 (3.86)
Salary	-2.42 (2.78)	-2.21 (2.54)	-4.56 (2.80)	-7.02* (3.66)
Salary Win-Lose	1.25 (3.31)	1.18 (3.06)	0.98 (3.37)	-0.18 (3.57)
Cuespread	0.09 *** (0.01)	0.10 *** (0.01)	0.10 *** (0.01)	0.10 *** (0.01)
Lagged Earnings		-9.28 *** (2.37)	-8.86 *** (2.37)	-8.94 *** (2.39)
Trait Anxiety			0.02 (0.12)	0.02 (0.12)
Female			6.81 *** (1.61)	6.80 *** (1.61)
Round	0.27 *** (0.09)	0.28 *** (0.09)	0.29 *** (0.09)	0.27 (0.20)
Constant	7.73 *** (2.64)	14.07 *** (3.54)	9.68 * (5.09)	9.90 * (5.50)
With treatment-round interactions	No	No	No	Yes
Observations	3690	3690	3390	3390
Participants	246	246	226	226
R²	0.093	0.114	0.130	0.131
p(PRWL = 0)	0.209	0.213	0.072	0.066
p(PRWL = T)	0.054	0.300	0.222	0.013
p(PRWL = SWL)	0.085	0.086	0.032	0.070
p(S = 0)	0.383	0.383	0.103	0.055
p(S = SWL)	0.182	0.178	0.047	0.057

Improvements in performance according to task difficulty

- Okay so tournaments do not do well overall
- But do tournaments perform relatively better when the task is easier?
- We can compare improvements in performance across the different tasks
- Many ways of doing this: we look at pairs of treatments and the differences in errors

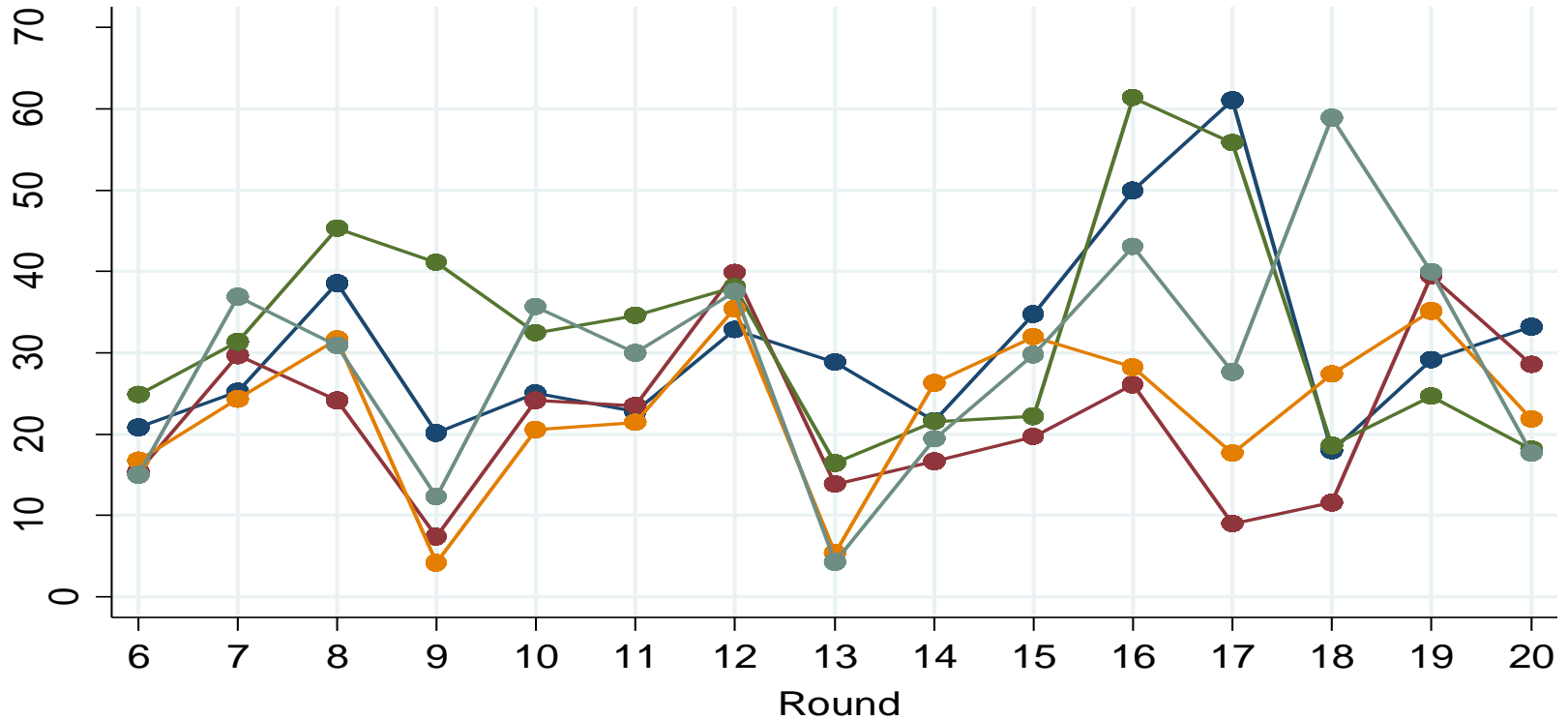
Piece Rate Win-Lose vs Tournament across task difficulty

Piece Rate Win-Lose errors - Tournament errors
Treatment Round Average Errors by Task

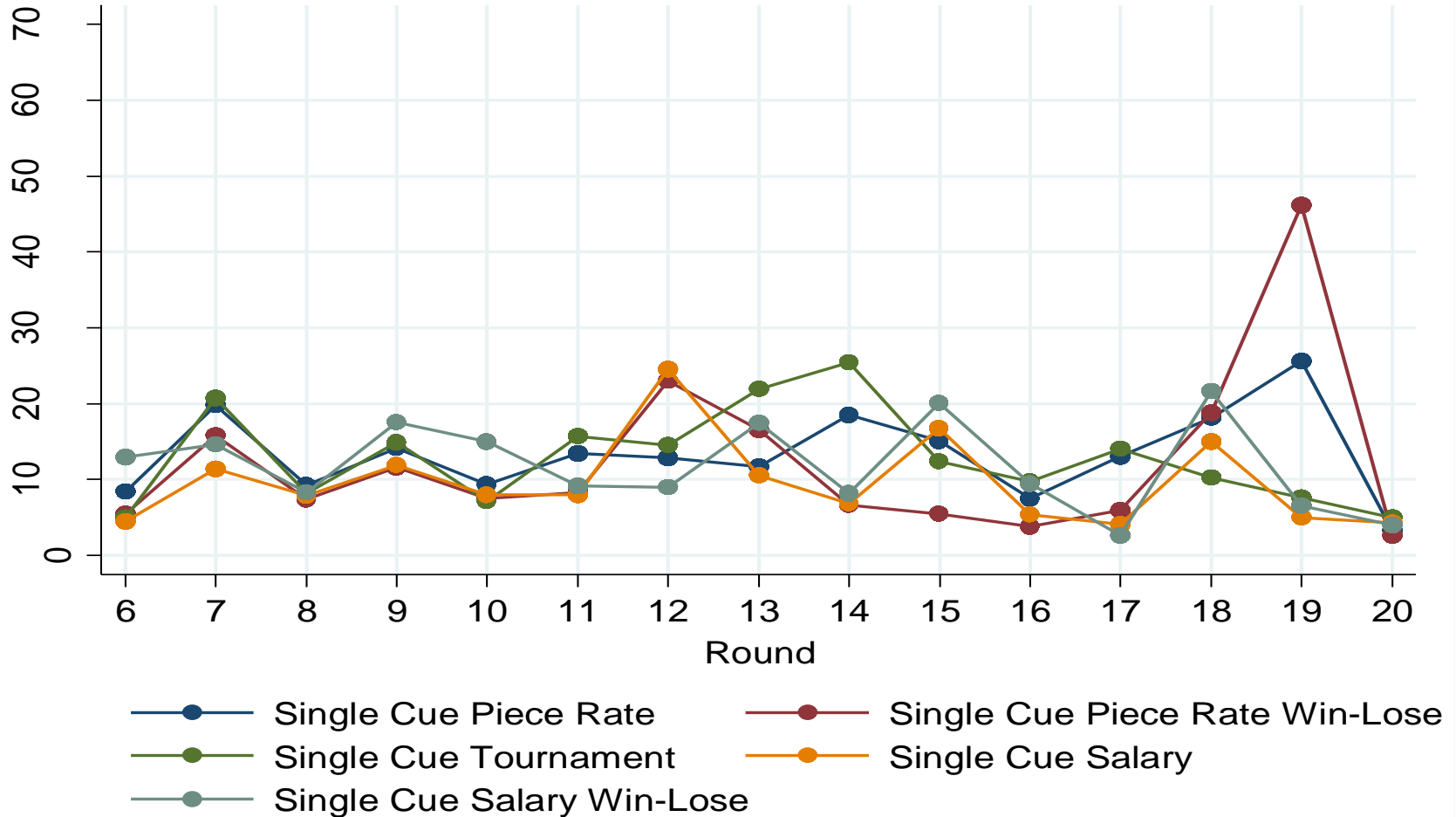


	Single Cue	Dual Cue	Ranksum
PRWL - T	-0.412 (3.135)	-6.707 (7.791)	z = 3.007 p = 0.0026

Learning over time?



Learning over time?



Dependent variable: Absolute forecast error
All Data – Partial Results for interaction terms

	Model 1 Rounds 5 - 20	Model 2 Rounds 11 - 20	Model 3 Rds 11-20 HIGH Perf (mean)	Model 4 Rds 11 - 20 LOW Perf (mean)
Round	0.176 (0.141)	0.490 ** (0.213)	0.038 (0.188)	0.652 (0.400)
Win-lose_round	0.112 (0.205)	-0.038 (0.370)	-0.239 (0.280)	0.438 (0.954)
Tournament_round	-0.294 (0.183)	-0.570 ** (0.282)	-0.541 ** (0.225)	-0.658 (0.553)
Salary_round	0.078 (0.183)	-0.041 (0.328)	-0.407 (0.287)	0.415 (0.662)
Salary Win-Lose_round	0.068 (0.212)	0.174 (0.318)	-0.160 (0.242)	0.244 (0.589)
Constant	6.494 (4.381)	5.287 (5.470)	5.616 (4.433)	7.124 (8.915)
Observations	5130	3420	1960	1460
Participants	342	342	196	146
R²	0.128	0.104	0.069	0.116

Dependent variable: Absolute forecast error
All Data – Partial Results for interaction terms

	Model 1 Rounds 5 - 20	Model 2 Rounds 11 - 20	Model 3 Rds 11-20 HIGH Perf	Model 4 Rds 11 - 20 LOW Perf
Round	0.176 (0.141)	0.490** (0.213)	-0.421*** (0.145)	0.644** (0.294)
Win-lose_round	0.112 (0.205)	-0.038 (0.370)	-0.402 (0.254)	0.210 (0.547)
Tournament_round	-0.294 (0.183)	-0.570** (0.282)	-0.429* (0.244)	-0.722* (0.397)
Salary_round	0.078 (0.183)	-0.041 (0.328)	0.055 (0.260)	-0.062 (0.514)
Salary Win-Lose_round	0.068 (0.212)	0.174 (0.318)	-0.198 (0.203)	0.305 (0.443)
Constant	6.494 (4.381)	5.287 (5.470)	10.268 *** (3.696)	7.217 (7.048)
Observations	5130	3420	1160	2260
Participants	342	342	116	226
R²	0.128	0.104	0.058	0.116

Accuracy of forecasts

Dep Var: Forecasts	Piece Rate	Win Lose	Tournament	Salary	Salary win-Lose
Cue A	0.34*** (0.02)	0.32*** (0.01)	0.30*** (0.02)	0.33*** (0.02)	0.36*** (0.02)
Cue B	0.63*** (0.02)	0.65*** (0.01)	0.62*** (0.02)	0.64*** (0.01)	0.61*** (0.02)
Constant	17.53*** (4.51)	15.13*** (1.96)	25.80*** (5.28)	17.53*** (2.00)	17.81*** (3.57)
Observations	1215	1155	1170	1140	960
Participants	81	77	78	76	64
R ²	0.873	0.906	0.846	0.910	0.860
Wald Chi ²	2018.92	11242.88	1465.87	8314.48	2321.44
p > chi ²	0.000	0.000	0.000	0.000	0.000
p(cue A = 0.3)	0.023	0.148	0.835	0.047	0.005
p(cue B = 0.7)	0.000	0.000	0.000	0.000	0.000
p(cons = 10)	0.095	0.009	0.003	0.000	0.029

High Performers

Dep Var: Forecasts	Piece Rate	Win Lose	Tournament	Salary	Salary win-Lose
Cue A	0.32 *** (0.02)	0.33 *** (0.01)	0.30 *** (0.03)	0.27 *** (0.01)	
Cue B	0.68 *** (0.01)	0.67 *** (0.01)	0.68 *** (0.02)	0.70 *** (0.01)	0.73 *** (0.01)
Constant	10.76 *** (1.97)	10.91 *** (2.36)	12.14 ** (5.05)	14.77 *** (2.84)	47.77 *** (1.90)
Observations	675	780	690	660	435
Participants	45	52	46	44	29
R ²	0.958	0.935	0.931	0.954	0.952
Wald Chi ²	7793.52	10004.43	2218.1	6774.69	5558.67
p > chi ²	0.000	0.000	0.000	0.000	0.0000
p(cue A = 0.3)	0.309	0.010	0.922	0.032	
p(cue B = 0.7)	0.153	0.019	0.232	0.835	0.007
p(cons = 10)	0.701	0.699	0.671	0.093	
p(cons = 55)					0.000

Low Performers

Dep Var: Forecasts	Piece Rate	Win Lose	Tournament	Salary	Salary win-Lose
Cue A	0.35 *** (0.02)	0.32 *** (0.03)	0.33 *** (0.03)	0.38 *** (0.02)	0.38 *** (0.02)
Cue B	0.59 *** (0.02)	0.63 *** (0.02)	0.55 *** (0.03)	0.58 *** (0.02)	0.56 *** (0.02)
Constant	22.45 *** (7.25)	19.03 *** (3.50)	33.44 *** (9.40)	19.83 *** (2.74)	21.46 *** (3.95)
Observations	540	375	480	480	525
Participants	36	25	32	32	35
R ²	0.813	0.863	0.778	0.881	0.837
Wald Chi ²	750.86	5967.71	517.41	5117.51	1930.74
p > chi ²	0.000	0.000	0.000	0.000	0.000
p(cue A = 0.3)	0.019	0.447	0.338	0.000	0.000
p(cue B = 0.7)	0.000	0.000	0.000	0.000	0.000
p(cons = 10)	0.086	0.010	0.013	0.000	0.004

Why does Salary Win-Lose do worse?

Data for rounds 6 - 20

	Salary Dual Cue	Salary Win- Lose Dual Cue	Salary Single Cue	Salary Win- Lose Single Cue
Cuespread	0.123*** (0.016)	0.154*** (0.021)	0.010*** (0.004)	0.025*** (0.007)
Round	0.692*** (0.234)	0.998*** (0.316)	-0.071 (0.0756)	-0.275** (0.128)
Trait Anxiety	NS	NS	0.158* (0.088)	0.361*** (0.092)
Female	NS	NS	NS	NS
Lagged Earnings	-10.95 (7.068)	-23.001** (9.168)	-2.365 (4.982)	-13.23* (7.51)
Constant	13.131 (8.274)	8.468 (17.668)	2.364 (7.0661)	8.105 (8.813)
Observations	465	405	600	480
Participants	31	27	40	32
R²	0.196	0.175	0.04	0.118

Concluding thoughts

- Across our two experiments by and large the treatments that perform better are “salary” and “win/lose” with salary doing better overall
- Part of the reason why tournaments do not perform well is because “low” performers fare especially poorly in this treatment
- Providing win/loss information in pay for performance schemes improves performance

Concluding thoughts

- However, providing win/loss information when payment is independent of performance actually makes things worse
- Tournament shows greater improvement in performance between dual and single cue tasks
- *This suggests that when a task is intellectually challenging tournaments may not do well but they might perform better if the task is more menial (?)*

Concluding thoughts

- Limited evidence of learning overall across different treatments
- But there is some evidence of learning in the tournament treatment particularly in the later rounds
- And this learning seems most pronounced for the “high” performers

Concluding thoughts

- Why does the salary treatment do well?
- Merlo and Schotter (1999)
 - ***Learn-while-you-earn*** and ***Learn-before-you-earn*** (LBYE)
 - *find that subjects do much better in the LBYE treatment where every single decision does not count for payment*
- *Why does Salary Win-Lose perform worse?*
 - *More myopic focus on per round earnings and winning/losing even when those do not matter?*
 - *Subjects feel “more controlled” when winning/losing information provided?*

Concluding thoughts

What is the aim?

- **Minimize aggregate errors**
 - *If pay independent of performance, then **Salary***
 - *If pay dependent on performance, then **Win-Lose***
- **Learning over time**
 - **Tournaments**
 - *especially for “highly skilled” workers*

Well, that's my story and I am sticking to it



Questions?