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### **Overview**

- **Research Question:** How individual investors learn about their trading talent.
- **Quantitative Result:** Learning about trading talent is about 7 times more sensitive to new signals than learning about stocks.
- Qualitative Result: A unifying framework for several documented puzzles
- Timing of stock switching
- Performance contingent attrition and trading intensity

### Abstract

Recent studies show evidence that investors learn about their trading abilities. This paper focuses on understanding how investors learn about their talent and proposes a unifying framework that explains many puzzling facts about individual equity investors. In my model, the investor forms subjective beliefs both about the expected return of the current stock-in-holding and about her trading talent represented by the expected return of the next replacement stock, and updates beliefs through learning with fading memory. I calibrate the memory decay parameters to individual trading records, and show that talent learning is about 7 times more sensitive to return signals than stock-in-holding learning. Consequently, the model indicates that stock switching always happens following good performance of the current stock because switching requires a sufficiently large wedge between expected returns of the replacement stock and the current stock to cover the fixed cost, which strongly predicts disposition effect in a learning perspective. This framework also accounts for the performance-contingent trading intensity and attrition, and explains why a negative shock leads to attrition when an investor has several years of experience, which is inconsistent with the decreasing-gain updating under standard Bayesian learning.

### Key Feature: Two Types of Learning

Bayesian learning about two subjects with constant-gain updating (Nagel and Xu, 2019).

- ► Trading talent
- Learning about ex-ante distribution of a draw of the subjective mean return  $\widetilde{\mu}_x$  of next replacement stock  $x_t$
- Normal density  $\pi_t(\widetilde{\mu}_x)$ : mean  $\widetilde{m}_{x,t}$  and variance  $\widetilde{s}_{r\,t}^2$





Figure 1. Estimation of Decay Parameters

### $\blacktriangleright$ Current stock $y_t$ held at t

- Learning about subjective mean return  $\widetilde{\mu}_{y}$  of the current stock  $y_{t}$
- Normal density  $\pi_t(\widetilde{\mu}_y)$ : mean  $\widetilde{m}_{y,t}$ and variance  $\widetilde{s}_{ut}^2$
- Temporary and specific to stock
- Why two types of learning? Attrition takes place  $\Rightarrow$  learning about more than stock
- ► Why only learn about mean? Under-diversified portfolio + relatively long horizon (not day traders) + simplify to demonstrate the mechanism
- Belief dynamics Weighted average of prior and new signal

$$\widetilde{m}_{x,t+1} = (1 - \nu_x) \, \widetilde{m}_y$$
$$\widetilde{m}_{y,t+1} = (1 - \nu) \, \widetilde{m}_y$$

• Memory decay parameter  $\nu$  and  $\nu_x$ Calibrated to individual holdings data.

 $\nu_x^{\star} = 0.144, \quad \nu^{\star} = 0.02$ 

## Subjective Learning of Trading Talent Theory and Evidence from Individual Investors in the U.S. Xindi He

### **Theory: Settings**

 $\nu_{x,t} + \nu_x r_{y,t+1}$  $= (1 - \nu) \tilde{m}_{y,t} + \nu r_{y,t+1}$ 

### Discrete time portfolio choice

- An investor chooses from a pool of stocks.
- Stock *i* log return  $r_{i,t+1} = \ln R_{i,t+1}$ : normal distribution

 $r_{i,t+1} = \mu_i + \varepsilon_{i,t+1}$ 

- $\varepsilon_{i,t+1} \sim \mathcal{N}(0,\sigma^2)$ : i.i.d. ( $\sigma^2$  is a known constant)
- $\mu_i$ : unobserved constant, not random, key parameter to learn, equivalently draw of  $\mu_i$
- Can only hold one stock at a time.
- Two decisions each period: (1) current vs. replacement; (2) investment consumption
- **Timing:** Within each period,
- Realized stock return  $r_{y,t}$  is observed
- Two subjective beliefs are updated
- Decisions on consumption, investment, and stock choice (replace or not) are made
- **Preferences:** Epstein-Zin utility function, i.e.,  $U_t = |(1 t_i)|$

### **Theory: Result**

- Methodology for analytical solution: Techniques in Campbell and Viceira (1999)
- Log-linearization of Euler equation, budget constraint and portfolio return
- Guess-and-verify method (Re-derived for the unit root state variable in my settings)
- Optimal investment and consumption:

$$a_{t}^{\star} = A_{0} + A_{1} \widetilde{m}_{y,t}$$

$$V_{t}^{\star} = \exp\left\{\frac{B_{0} - \frac{1}{\rho}\ln\left(1 - \beta\right)}{1 - \frac{1}{\rho}} + \frac{B_{1}}{1 - \frac{1}{\rho}}\widetilde{m}_{y,t} + \frac{B_{2}}{1 - \frac{1}{\rho}}\widetilde{m}_{y,t}^{2}\right\}$$
where Compare

Optimal switching rule: Compare

$$U_{t}(y_{t}) = W_{t} \exp\left\{\frac{B_{0} - \frac{1}{\rho}\ln(1-\beta)}{1 - \frac{1}{\rho}} + \frac{B_{1}}{1 - \frac{1}{\rho}}\widetilde{m}_{y_{t},t} + \frac{B_{2}}{1 - \frac{1}{\rho}}\widetilde{m}_{y_{t},t}^{2}\right\}$$
$$U_{t}(y_{x,t}) = (1-\theta)W_{t} \exp\left\{\frac{B_{0} - \frac{1}{\rho}\ln(1-\beta)}{1 - \frac{1}{\rho}} + \frac{B_{1}}{1 - \frac{1}{\rho}}\widetilde{m}_{x,t} + \frac{B_{2}}{1 - \frac{1}{\rho}}\widetilde{m}_{x,t}^{2}\right\}$$

and switch to the new stock  $y_{x,t}$  iff  $U_t(y_{x,t}) > U_t(y_t)$ 

**Properties:** Monotonicity  $(A_1 > 0)$  and Convexity  $(\frac{B_2}{1-1} > 0)$ 



Figure 2. Value Function

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$$\beta) C_t^{1-\rho} + \beta \left( \widetilde{\mathbb{E}}_t \left[ U_{t+1}^{1-\gamma} \right] \right)^{\frac{1-\rho}{1-\gamma}} \right]^{\frac{1}{1-\rho}}$$

- 1996
- ► Filter:
- Long-only investors that start trading after February 1991
- Quit at most once
- Have records of at least 3 consecutive months
- All stocks in holding matched to CRSP monthly database
- ► Total number of accounts: 7817

### Timing of stock switching

• Regression:  $switch_{i,t+1} = \lambda_0 + \lambda_1$ 



enough to cover the transaction cost.

### Attrition and performance

• Regression:  $Y_i = \phi_0 + \phi_1 \widetilde{m}_{i,x} + \phi_2 \widetilde{m}_{i,y} + \phi_3 num \ switch_i + \epsilon_i$ 

	(1)	(2)	(3)	(4)	(5)	(6)	
_	Dependent Variable: $not \ quit_i$			Dependent Variable: $\overline{return_i}$			
$ar{ ilde{m}}_{i,x}$ –	2.934***	2.477***	2.717***			1.5047***	
	(0.299)	(0.428)	(0.299)			(0.0340)	
$ar{ ilde{m}}_{i,y}$		1.356			$2.0317^{***}$		
		(0.955)			(0.0752)		
$num \ switch_i$			0.048***	0.0020***	$-0.0016^{***}$	$-0.0009^{***}$	
			(0.007)	(0.0003)	(0.004)	(0.002)	
Obs.	7,817	7,817	7,817	7,817	7,817	7,817	

- Takeaway:
- dropping out of the market as implied by the model.
- investors. The Journal of Finance, 55(2):773-806, 2000.
- Economics, 114(2):433-495, 1999.
- [3] Stefan Nagel and Xu Zhengyang. Asset pricing with fading memory. *Review of Financial Studies*, 2021.

### **Evidence:** Data

▶ Trading records data of individual investors in the U.S. between February 1991 and November

### **Evidence: Regressions**

$\times ($	$\widetilde{m}_{i,x,t}$ -	$-\widetilde{m}_{i,y,j}$	$t) + \lambda'_{z}$	${}_{3}^{\prime} X_{i,t} +$	$-u_i + \varepsilon_i$	$,t{+}1$
(2)	(3)	(4)	(5)	(6)	(7)	(8)
	Dep	endent Var	iable: <i>swit</i>	$ch_{i,t+1}$		
07*** 026)	$0.051^{***}$ (0.013) $0.318^{***}$ (0.052)	$0.110^{***}$ (0.032) -0.026 (0.099)	$0.101^{***}$ (0.015) $-0.001^{*}$ (0.001)	0.080*** (0.025) 0.003*** (0.001)	$0.086^{***}$ (0.016) $0.381^{***}$ (0.058) $-0.003^{***}$ (0.001) $0.004^{*}$	$\begin{array}{c} 0.084^{***} \\ (0.028) \\ -0.090 \\ (0.111) \\ 0.004^{***} \\ (0.001) \\ 0.006^{**} \end{array}$
					(0.003)	(0.002)

• Takeaway: As implied by the model, stock switching is more likely to happen when the wedge between the subjective beliefs about trading talent and current stock is large

• Higher subjective mean of talent distribution pushes the investor further away from • More frequent stock switching hurts investment performance (Barber and Odean, 2000).

### References

[1] Brad M Barber and Terrance Odean. Trading is hazardous to your wealth: The common stock investment performance of individual

[2] John Y Campbell and Luis M Viceira. Consumption and portfolio decisions when expected returns are time varying. *Quarterly Journal of* 

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