

# Hope, Fear, and Aspiration

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Based on the joint work with Prof. Xunyu Zhou in Oxford

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  - Lopes' SP/A theory (Lopes 1987, Lopes and Oden 1999)

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- Inspired by these models of choice, we propose a new portfolio choice model featuring *hope*, *fear* and *aspiration*
- This model is based on SP/A theory and rank-dependent utility
- Analytical solutions are found and the impact of hope, fear, and aspiration is studied

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- Hope and fear coexist. Hope is optimism over extremely satisfactory situations and fear is pessimism over really poor situations
- When evaluating a random prospect, hope makes the agent overweight the probability of best outcomes, while fear makes him overweight the probability of worst outcomes (rank-dependent utility theory and prospect theory)

# Hope and Fear (Con'd)

- We appeal to rank-dependent utility to model hope and fear

$$\begin{aligned}V(X) &:= \int_0^{+\infty} u(x) d[-w(1 - F_X(x))] \\ &= \int_0^{+\infty} u(x) w'(1 - F_X(x)) dF_X(x)\end{aligned}$$

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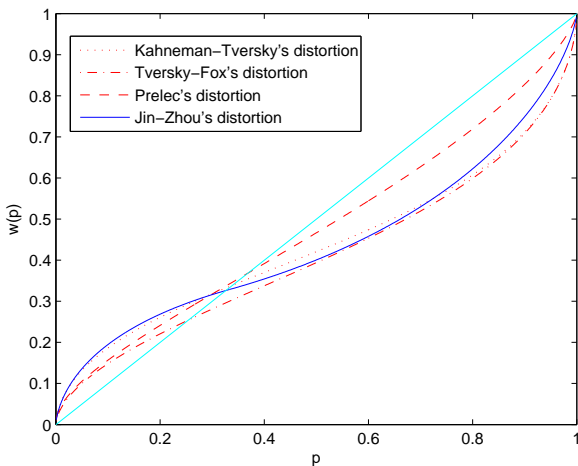
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- $u(\cdot)$  is typically concave
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- $w(\cdot)$ , *probability distortion function*, is usually reverse-S shaped, capturing hope and fear simultaneously

# Distortion Functions



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  - Outside influence
- Aspiration is modeled by the constraint

$$P(X \geq A) \geq \alpha$$

where  $A$  is *aspiration level*,  $\alpha$  is *confidence level*

# Literature

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- He and Zhou (2010) developed a general method to deal with the distortion function in the context of continuous-time portfolio choice, and applied it to Yaari's model
- Carlier and Dana (2009) considered the portfolio selection problem with rank-dependent utility

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- Martingale approach applied
- A static problem

# Portfolio Choice Model

- *HF/A portfolio choice model*

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 \text{Max}_X & \int_0^\infty u(x) d[-w(1 - F_X(x))] \\
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- Quantile formulation (Schied 2004, Carlier and Dana 2006, Jin and Zhou 2008, He and Zhou 2010)

# Quantile Formulation

- The HF/A portfolio choice model

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$$V(X) = \int_0^1 u(F_X^{-1}(z)) w'(1 - z) dz$$

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- Budget constraint (by Hardy-Littlewood inequality)

$$\int_0^1 F_\rho^{-1}(1 - z) F_X^{-1}(z) dz \leq x_0$$

# Quantile Formulation (Cont'd)

- Let  $G(\cdot) = F_X^{-1}(\cdot)$ , we derive the quantile formulation

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- If  $G^*(\cdot)$  is an optimal solution to the quantile formulation, then  $X^* := G^*(1 - F_\rho(\rho))$  is an optimal terminal wealth

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- Explicit solutions can be found

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- Carlier and Dana (2009) derive a similar result

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- Higher hope index makes the agent gamble more on the best scenarios so that he can get higher payoff when these scenarios really happen.

# Lottery-likeness Index

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- The lottery-likeness index measures how the optimal terminal is like a lottery, serving as an indicator of how aggressive the agent is.
- **Theorem:**  $\mathcal{L}(A)$  can be computed explicitly and is increasing w.r.t the aspiration level  $A$

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- Jin-Zhou distortion

$$w(z) = \begin{cases} ke^{b\mu_\rho + (a+b)\sigma_\rho\Phi^{-1}(1-z_0) + \frac{(a\sigma_\rho)^2}{2}}\Phi(\Phi^{-1}(z) + a\sigma_\rho), & z \leq 1 - z_0, \\ A + ke^{b\mu_\rho + \frac{(b\sigma_\rho)^2}{2}}\Phi(\Phi^{-1}(z) - b\sigma_\rho), & z \geq 1 - z_0 \end{cases}$$

where  $k$  is determined by  $a, b, z_0$ .  $b$  measures fear and  $a$  measures hope. Choose  $a = 3$ ,  $b = 2.2$  and  $z_0 = \frac{2}{3}$ .

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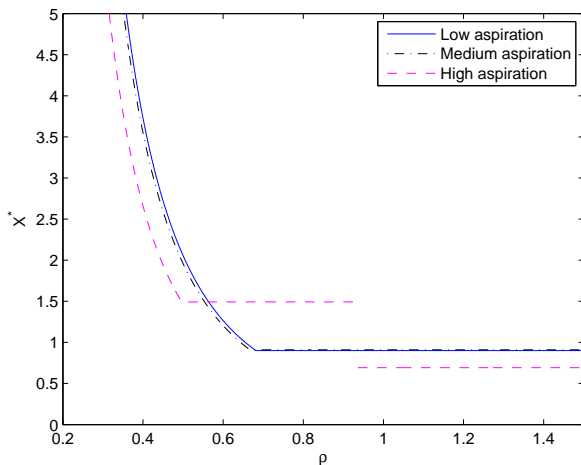
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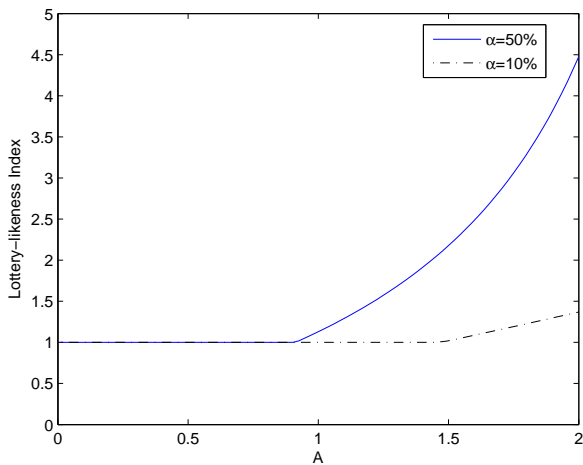
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- Initial budget  $x_0 = 1$

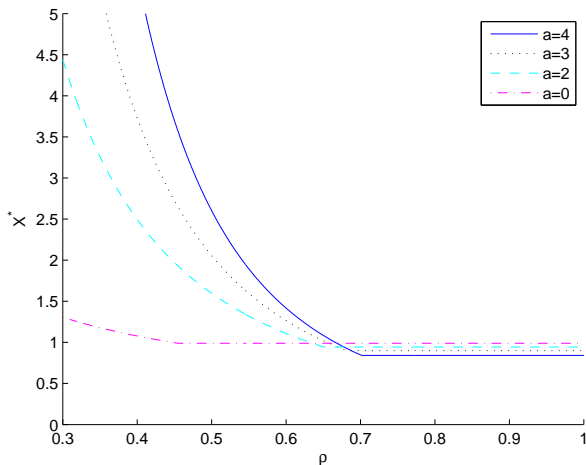
# Optimal Solution with Different Aspiration Levels



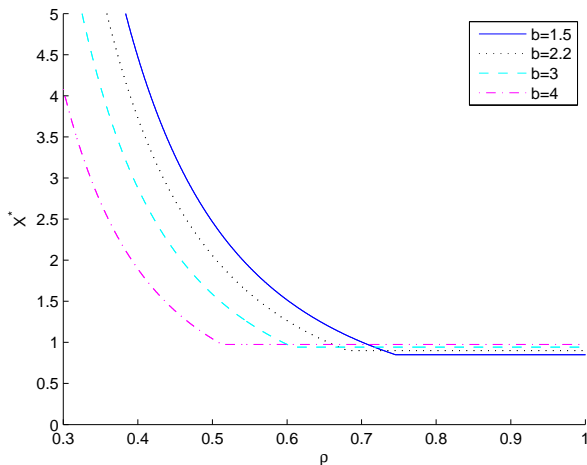
# Lottery-likeness Index with Different Confidence Levels



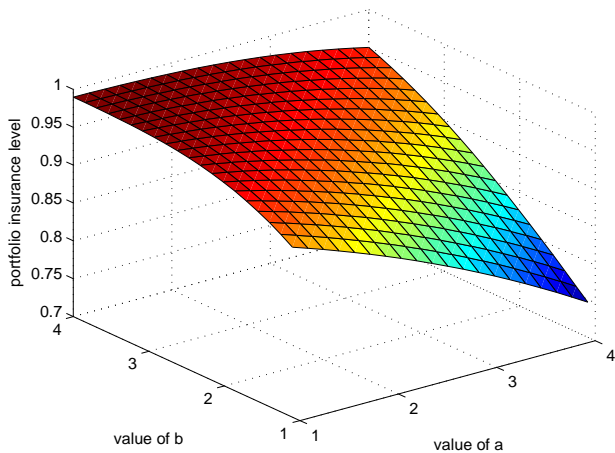
# Optimal Solutions with Different Levels of Hope



# Optimal Solutions with Different Levels of Fear



# Portfolio Insurance Level



# Dynamic Portfolio

- Optimal dynamic portfolio (the dollar amount in stock at each time  $t$ )

$$\pi(t) = \Delta(t, \rho(t)) \frac{1}{\eta} (\sigma^\top)^{-1} \theta X(t), \quad 0 \leq t < T.$$



# Dynamic Portfolio

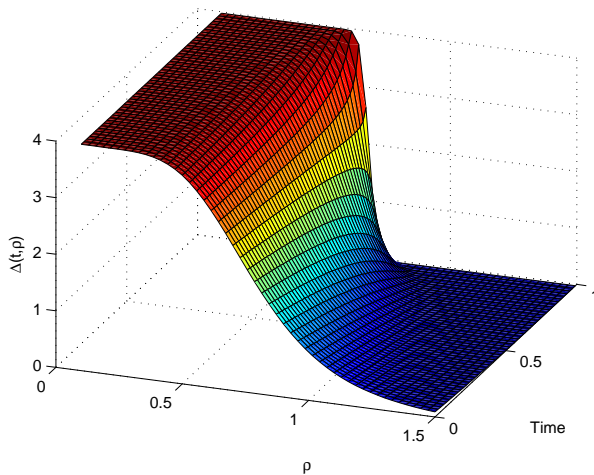
- Optimal dynamic portfolio (the dollar amount in stock at each time  $t$ )

$$\pi(t) = \Delta(t, \rho(t)) \frac{1}{\eta} (\sigma^\top)^{-1} \theta X(t), \quad 0 \leq t < T.$$

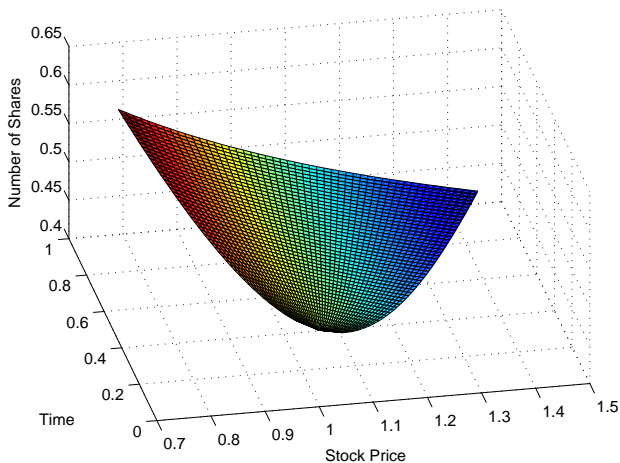
- $\Delta(t, \rho)$  is strictly increasing w.r.t  $\rho$  and

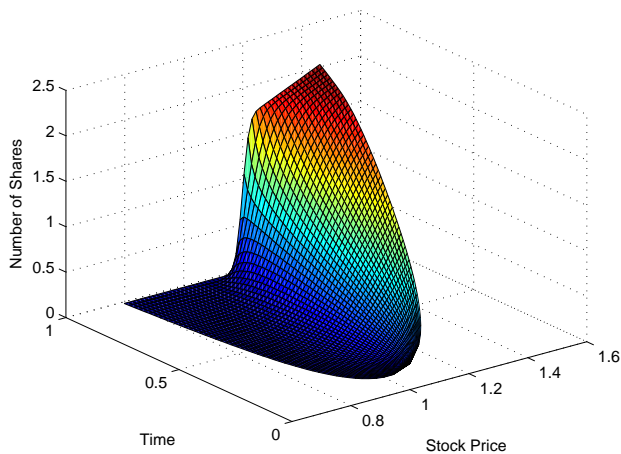
$$\lim_{\rho \downarrow 0} \Delta(t, \rho) = a + 1, \quad \lim_{\rho \uparrow \infty} \Delta(t, \rho) = 0$$

# Deviation from EUT



# Number of Shares in Stock: EUT with $\eta = 5$



Number of Shares in Stock: HF/A with  $\eta = 5$ 

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- Strong fear leads to portfolio insurance
- Hope makes investors gamble more on good situations
- High aspiration produces aggressive strategies