# **EE365: Linear Quadratic Trading Example**

#### Linear quadratic trading: Dynamics

- $x_{t+1} = f_t(x_t, u_t, \rho_t) = \mathbf{diag}(\rho_t)(x_t + u_t)$
- $> x_t \in \mathbb{R}^n$  is dollar amount of holding in n assets
- $(x_t)_i < 0$  means short position in asset i in period t
- $lackbox{} u_t \in \mathbb{R}^n$  is dollar amount of each asset bought at beginning of period t
- $(u_t)_i < 0$  means asset i is sold in period t
- $lackbox{ } x_t^+ = x_t + u_t$  is post-trade portfolio
- $ho_t \in \mathbb{R}^n_{++}$  is (random) return of assets over period (t,t+1]
- ▶ returns independent, with  $\mathbf{E} \, \rho_t = \overline{\rho}_t$ ,  $\mathbf{E} \, \rho_t \rho_t^\mathsf{T} = \Sigma_t$

## Linear quadratic trading: Stage cost

stage cost for  $t = 0, \dots, T-1$  is (convex quadratic)

$$g_t(x, u) = \mathbf{1}^{\mathsf{T}} u + \frac{1}{2} (\kappa_t^{\mathsf{T}} u^2 + \gamma (x + u)^{\mathsf{T}} Q_t(x + u))$$

with  $Q_t > 0$ 

- ▶ first term is gross cash in
- ightharpoonup second term is quadratic transaction cost (square is elementwise;  $\kappa_t > 0$ )
- ▶ third term is risk (variance of post-trade portfolio for  $Q_t = \Sigma_t \overline{\rho}_t \overline{\rho}_t^\mathsf{T}$ )
- $ightharpoonup \gamma > 0$  is risk aversion parameter
- minimizing total stage cost equivalent to maximizing (risk-penalized) net cash taken from portfolio

#### Linear quadratic trading: Terminal cost

- ▶ terminal cost:  $g_T(x) = -\mathbf{1}^\mathsf{T} x + \frac{1}{2} \kappa_T^\mathsf{T} x^2$ ,  $\kappa_T > 0$
- ▶ this is net cash in if we close out (liquidate) final positions, with quadratic transaction cost

## Linear quadratic trading: DP

value functions quadratic (including linear and constant terms):

$$v_t(x) = \frac{1}{2}(x^{\mathsf{T}} P_t x + 2q_t^{\mathsf{T}} x + r_t)$$

▶ we'll need formula

$$\mathbf{E}(\mathbf{diag}(\rho_t)P\,\mathbf{diag}(\rho_t)) = P \circ \Sigma_t$$

where o is Hadamard (element-wise) product

optimal expected tail cost

$$\mathbf{E} \, v_{t+1}(f_t(x, u, \rho_t)) = \mathbf{E} \, v_{t+1}(\mathbf{diag}(\rho_t) x^+) = \frac{1}{2} ((x^+)^\mathsf{T} P_{t+1} \circ \Sigma_t x^+ + 2q_{t+1}^\mathsf{T} \, \mathbf{diag}(\overline{\rho}_t) x^+ + r_{t+1})$$

## Linear quadratic trading: DP

- ▶  $P_T = \mathbf{diag}(\kappa_T), q_T = -1, r_T = 0$
- recall  $v_t(x) = \min_u \mathbf{E} \left( g_t(x, u) + v_{t+1}(\mathbf{diag}(\rho_t)(x+u)) \right)$
- ▶ for t = T 1, ..., 0 we minimize over u to get optimal policy:

$$\mu_t(x) = \operatorname{argmin}_u \left( u^{\mathsf{T}} (S_{t+1} + \mathbf{diag}(\kappa_t)) u + 2(S_{t+1}x + s_{t+1} + \mathbf{1})^{\mathsf{T}} u \right)$$

$$= -(S_{t+1} + \mathbf{diag}(\kappa_t))^{-1} (S_{t+1}x + s_{t+1} + \mathbf{1})$$

$$= K_t x + l_t$$

where

$$S_{t+1} = P_{t+1} \circ \Sigma_t + \gamma Q_t, \qquad s_{t+1} = \overline{\rho}_t \circ q_{t+1}$$

• using  $u = K_t x + l_t$  we then have

$$v_{t}(x) = \frac{1}{2} \begin{bmatrix} x \\ 1 \end{bmatrix}^{\mathsf{T}} \begin{bmatrix} S_{t+1}(I+K_{t}) & s_{t+1} + S_{t+1}l_{t} \\ s_{t+1}^{\mathsf{T}} + l_{t}^{\mathsf{T}}S_{t+1} & r_{t+1} + (s_{t+1}+\mathbf{1})^{\mathsf{T}}l_{t} \end{bmatrix} \begin{bmatrix} x \\ 1 \end{bmatrix}$$

#### Linear quadratic trading: value iteration

- ightharpoonup set  $P_T = \mathbf{diag}(\kappa_T)$ ,  $q_T = -1$ ,  $r_T = 0$
- ▶ for t = T 1, ..., 0

$$K_{t} = -(S_{t+1} + \operatorname{diag}(\kappa_{t}))^{-1} S_{t+1}$$

$$l_{t} = -(S_{t+1} + \operatorname{diag}(\kappa_{t}))^{-1} (s_{t+1} + 1)$$

$$P_{t} = S_{t+1} (I + K_{t})$$

$$q_{t} = s_{t+1} + S_{t+1} l_{t}$$

$$r_{t} = r_{t+1} + (s_{t+1} + 1)^{\mathsf{T}} l_{t}$$

where

$$S_{t+1} = P_{t+1} \circ \Sigma_t + \gamma Q_t, \qquad s_{t+1} = \overline{\rho}_t \circ q_{t+1}$$

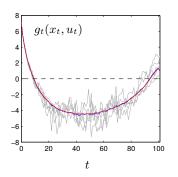
- optimal policy:  $\mu_t^{\star}(x) = K_t x + l_t$
- ► can write as  $\mu_t^*(x) = K_t(x x_t^{\text{tar}}), \quad x_t^{\text{tar}} = -K_t^{-1}l_t = -S_{t+1}^{-1}(s_{t+1} + 1)$

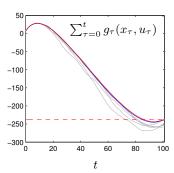
#### Linear quadratic trading: Numerical instance

- ightharpoonup n=30 assets over T=100 time-steps
- ightharpoonup initial portfolio  $x_0 = 0$
- $ightharpoonup \overline{\rho}_t = \overline{\rho}, \ \Sigma_t = \Sigma \ \text{for} \ t = 0, \dots, T-1$
- $ightharpoonup Q_t = \Sigma \overline{\rho}\overline{\rho}^{\mathsf{T}} \text{ for } t = 0, \dots, T-1$
- lacktriangle asset returns log-normal, expected returns range over  $\pm 3\%$  per period
- $\blacktriangleright$  asset return standard deviations range from 0.4% to 9.8%
- ightharpoonup asset correlations range from -0.3 to 0.8

# Linear quadratic trading: Numerical instance

- ightharpoonup N=100 Monte Carlo simulations
- ▶  $J^* = v_0(x_0) = -237.5$  (Monte Carlo estimate: -238.4)
- $\blacktriangleright$  exact (red), MC estimate (blue), and samples (gray);  $J^*$  red dashed





# Linear quadratic trading: Numerical instance

we define  $x_{T+1} = 0$ , i.e., we close out the position during period T

