Risk Sensitive Portfolio Optimization - Dynamical Models

Łukasz Stettner, Institute of Mathematics Polish Acad. Sci.

October 11, 2008

How to model asset prices?

```
discrete or continuous time discrete time:
\frac{S_i(t+1)}{S_i(t)} = 1 + \zeta_i(z(t), \xi(t+1))
\frac{S_i(t+s)}{S_i(t)} = 1 + \zeta_i(t,s,z(.),\xi(.))
\xi(.) independent (increments) "noise" process
stationarity
(1 + \zeta_i(t, s, z(.), \xi(.))(\omega)) = (1 + \zeta_i(s, z(.), \xi(.))(\theta_t\omega)) is a
multiplicative functional
1 + \zeta_i(t + s, z(.), \xi(.))(\omega) =
(1 + \zeta_i(t, z(.), \xi(.))(\omega))(1 + \zeta_i(s, z(.), \xi(.))(\theta_t\omega)).
1 + \zeta_i(s, z(.), \xi) := e^{X_i(s)}
dX(t) = \alpha(z(t))dt + \sigma(z(t))dB(t) + \int_{R^d} \gamma(z(t), u)\tilde{N}(dt, du)
X(0) = 0 and \tilde{N} is a compensated Poisson measure
z(t) Markov process (ergodic) of economic factors
key problem: Markovianity (S(t), z(t)) is a Markov process
problem: replace (B(t)) by a more general Gaussian (e.g. fractional
Gaussian) processes
```

Wealth process and portfolio strategies

no transaction costs

$$N(t) = (N_1(t), \dots, N_d(t))^T$$
 number of assets in portfolio at time t .

$$V(t)$$
 wealth process at time t

$$\sum_{i=1}^{d} N_i(t) S_i(t) = V(t)$$

$$\pi_i(t) = \frac{N_i(t)S_i(t)}{V(t)}$$

Wealth and portfolio dynamics:

$$V(t+s) = V(t)\pi(t) \cdot (1 + \zeta(t,s,z(.),\xi))$$

$$\pi(t+s) = g(\pi(t) \diamond e^{X(t+s)-X(t)})$$

where

$$\pi(t) \diamond e^{X(t+s)-X(t)} = (\pi_i(t)(1+\zeta_i(t,s,z(.),\xi)) = (\chi_i(t)) + \chi_i(t+s) - \chi_i(t)$$

$$(\pi_i(t)e^{X_i(t+s)-X_i(t)})$$

$$g_i(\pi) = \frac{\pi_i}{\sum_{j=1}^m \pi_j}$$

Terminal utility maximization - discrete time

```
E_{z,V}\left\{\sum_{t=0}^{T-1}f(V(t))+h(V(T))
ight\}
ightarrow \mathsf{max}
strategy \Pi = (\pi(t)), e.g. h(V) = V - \lambda(V - z)^{-1}
Bellman equations
w_T(z, V) = h(V)
w_{\tau_{-1}}(z, V) = f(V) + \sup_{\pi} E_{z, V} \{h(V\pi \cdot (1 + \zeta(1, z, \xi)))\}
w_n(z, V) = f(V) + \sup_{\pi} E_{z,V} \{ w_{n+1}(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \}
w_0(z, V) = f(V) + \sup_{\pi} E_{z, V} \{ w_1(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \}
Theorem
w_0(z, V) = \sup_{\Pi} E_{z, V} \left\{ \sum_{t=0}^{T-1} f(V(t)) + h(V(T)) \right\}
\hat{\pi}(t) = \tilde{\pi}_t(z(t), V(t))
```

Other functionals

```
infinite horizon models:
discounted reward:
w^{\beta}(z,V) = \sup_{\Pi} E_{z,V} \left\{ \sum_{t=0}^{\infty} \beta^t f(V(t)) \right\}
discounted Bellman equation:
w^{\beta}(z, V) = f(V) + \beta \sup_{\pi} E_{z, V} \{ w^{\beta}(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \}
long time horizon:
w(z, V) = \sup_{\Pi} \liminf_{T \to \infty} \frac{1}{T} E_{z, V} \left\{ \sum_{t=0}^{T-1} f(V(t)) \right\}
long time Bellman equations:
r(z, V) = f(V) - \gamma + \sup_{\pi} E_{z, V} \{ r(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \}
Theorem: w(z, V) = \gamma
example: growth optimal portfolio - risk neutral case
\liminf_{T\to\infty} \frac{1}{T} E_{z,V} \left\{ \ln(V(T)) \right\} = J_0(\Pi)
```

Risk sensitive functionals

```
finite horizon
w(z, V) =
\sup_{\Pi} \frac{-1}{\lambda} \ln E_{z,V} \left\{ \exp \left\{ -\lambda \left( \sum_{t=0}^{T-1} f(V(t)) + h(V(T)) \right) \right\} \right\}
w(z, V) = \sup_{\Pi} \frac{-1}{\lambda} \ln E_{z, V} \{ \exp \{ -\lambda \ln(V(T)) \} \} =
= \sup_{\Gamma} \frac{-1}{N} \ln E_{z,V} \{ \{ (V(T))^{-\lambda} \} \}
Bellman equations:
r_{\tau}(z,V)=h(V)
r_{T-1}(z, V) =
f(V) + \inf_{\pi} \ln E_{z,V} \{ \exp \{ r_T(z(1), V_{\pi} \cdot (1 + \zeta(z, \xi(1))) \} \}
r_n(z, V) =
f(V) + \inf_{\pi} \ln E_{z,V} \{ \exp \{ r_{n+1}(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \} \}
r_0(z, V) = f(V) + \inf_{\pi} \ln E_{z, V} \{ \exp \{ r_1(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \} \}
w(z, V) = \frac{-1}{2} r_0(z, V)
```

Long Run Risk Sensitive Functionals

```
w(z, V) =
\sup_{\Pi} \liminf_{T \to \infty} \frac{-1}{\lambda T} \ln E_{z,V} \left\{ \exp \left\{ -\lambda \left( \sum_{t=0}^{T-1} f(V(t)) + h(V(T)) \right) \right\} \right\}
w(z, V) = \sup_{\Gamma} \liminf_{T \to \infty} \frac{-1}{2T} \ln E_{z, V} \left\{ \exp \left\{ -\lambda \ln(V(T)) \right\} \right\} =
\sup_{\Pi} J_{\lambda}(\Pi)
Bellman equation:
r(z, V) =
\lambda f(V) - \gamma_{\lambda} \lambda + \inf_{\pi} \ln E_{z,V} \left\{ \exp \left\{ r(z(1), V\pi \cdot (1 + \zeta(z, \xi(1))) \right\} \right\}
w(z, V) = \gamma_{\lambda}
\lim_{\lambda \to 0} J_{\lambda}(\Pi) = J_{0}(\Pi)
\lim_{\lambda \to 0} \gamma_{\lambda} = \gamma
```

Problems with Transaction Costs

proportional transaction costs

$$S_i(t), t \geq 0, i = 1, 2, ..., d$$

V(t), $V^-(t)$ wealth processes after and before transaction at time t, $N_i(t)$, $N_i^-(t)$ number of assets at time t in portfolio after and before transaction

transaction cost at t (assuming we make a transaction

$$N(t) \neq N^-(t)$$

$$C(N(t), N^-(t), S(t)) =$$

$$\sum_{i=1}^{d} c_{1i}((N_i(t)-N_i^-(t))S_i(t))^+ + \sum_{i=1}^{d} c_{2i}((N_i(t)-N_i^-(t))S_i(t))^-$$

$$C(N, N^-, S) = \sum_{i=1}^d c_{1i}((N_i - N_i^-)S_i)^+ + \sum_{i=1}^d c_{2i}((N_i - N_i^-)S_i)^-$$

selffinancing property $V^- = V + C(N, N^-, S)$.

$$N^- \cdot S = N \cdot S + C(N, N^-, S)$$

 $V^{-}(t)$ the wealth before transaction, V(t) after transaction,

$$\pi_{i}^{-}(t) = \frac{N_{i}^{-}(t)S_{i}(t)}{V^{-}(t)}, \ \pi_{i}(t) = \frac{N_{i}(t)S_{i}(t)}{V(t)}, \ \pi(t) \in \mathcal{S}$$

$$V^{-}(t) = V(t) + V^{-}(t)c(\pi' - \pi^{-}(t))$$
, where

Problems with Transaction Costs (cont.)

```
 \begin{aligned} \pi' &\in \mathcal{S}^0 \text{ is s.t. } \pi(t) = g(\pi') \text{ and } \\ c(\nu) &= \sum_{i=1}^d c_{1i}(\nu_i)^+ + \sum_{i=1}^d c_{2i}(\nu_i)^- \\ &\sum \pi'_i + c(\pi' - \pi^-(t)) = 1 \text{ and } \pi(t) = g(\pi'), \ g_i(\pi) = \frac{\pi_i}{\sum_{j=1}^m \pi_j} \\ \text{Lemma 1. Given } \pi, \bar{\pi} &\in \mathcal{S} \text{ there is a continuous } e(\pi, \bar{\pi}) \in (0, 1] \text{ s.t. } \\ F^{\pi, \bar{\pi}}(e(\pi, \bar{\pi})) &= 1 \text{ with } F^{\pi, \bar{\pi}}(\delta) = \delta + c(\delta \bar{\pi} - \pi) \\ V^{\Pi, \nu}(t) &= e(\pi^-(t), \pi(t)) V^{\Pi, \nu -}(t) \\ \text{If there are no transactions in } (s, s + t) \\ V^{\Pi, \nu}(t + s)^- &= V^{\Pi, \nu}(t) \pi(t) \cdot e^{X(t+s) - X(s)} \\ \pi^-(t + s) &= g(\pi(t) \diamond e^{X(t+s) - X_i(t)}) \\ \pi(t) &\diamond e^{X(t+s) - X(s)} &= (\pi_i(t) e^{X_i(t+s) - X_i(t)}) \end{aligned}
```

Asymtotics of $\pi(t)$ (without transactions)

```
1 + \zeta_i(s, z(.), \xi) := e^{X_i(s)}
dX(t) = \alpha(z(t))dt + \sigma(z(t))dB(t) + \int_{R^d} \gamma(z(t), u) \tilde{N}(dt, du)
X(0) = 0
by the Law of Large Numbers
\lim_{t \to \infty} \frac{1}{t} X_i(t) = \lim_{t \to \infty} \frac{1}{t} \int_0^t \alpha_i(z(s)) ds = \int \alpha_i(z) \mu(dz) = r_i
\mu \text{ a unique invariant measure for } (z(t))
If \exists_k r_k > r_i \text{ for } i \neq k \text{ then}
\pi(t) \to \pi(\infty) = \delta_k, \text{ provided } \pi_k(0) > 0
```

Asymptotics of the portfolio growth (without transactions)

$$\begin{split} &\ln(\pi \cdot e^{X(n)}) = \sum_{t=0}^{n-1} \ln \pi(t) \cdot e^{X(t+1)-X(t)} \; (X(0)=0) \\ &\lim_{n \to \infty} \frac{1}{n} \left(\ln(\pi \cdot e^{X(n)}) - \sum_{t=0}^{n-1} w(\pi(t), z(t)) \right) = 0 \\ &w(\pi, z) = E_z \left[\ln \pi \cdot e^{X(1)} \right] \\ &\lim_{n \to \infty} \frac{1}{n} \sum_{t=0}^{n-1} w(\pi(t), z(t)) = \int w(\pi(\infty), z) \mu(dz) \\ &\lim_{n \to \infty} \frac{1}{n} \ln(\pi \cdot e^{X(n)}) = \int w(\pi(\infty), z) \mu(dz) \\ &\pi \mapsto w(\pi, z) \text{ has at most one interior supremum point in } \mathcal{S}, \text{ can we get better than } \int w(\pi(\infty), z) \mu(dz) \; ? \end{split}$$

Continuous time models

```
feasible control impulsive control \Pi=(\tau_i,\pi_{\tau_i}) at time \tau_i we choose portfolio \pi_{\tau_i} for models with fixed plus proportional transaction costs the class of impulse controls is optimal how to solve such problems: quasi variational inequalities
```

Partial observation models in finance

```
\frac{S_i(t+1)}{S_i(t)} = 1 + \zeta_i(z(t), \xi(t+1))
\frac{S_i(t+s)}{S_i(t)} = 1 + \zeta_i(t, s, z(.), \xi(.))
we do not observe (z(t))
can we recover z(t) from \{S(u); u \leq t\}?
how to calculate strategies: discretization (Runggaldier Stettner
1993), quantization,
problem with computational complexity
General question: how to calculate strategies for discrete or
continuous time models?
in the worst cases the main tool is Monte Carlo Method
```