Exotic Options in Multiple Priors Models

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Outline



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General Framework

Exotic Options in Multiple priors Models

Conclusions

Motivation

Classical Exercise Problem for American Options



Conclusions

American options:

- The right to buy or sell an underlying S at any time prior to maturity T subject to a contract
- Realizing the profit $A(t, (S_s)_{s \le t})$ when exercised at t

Problem of the buyer:

Exercise the option optimally choosing a strategy that maximizes the expected reward of the option, i.e. choose a stopping time τ^* that maximizes

 $\mathbb{E}^{P}((A(\tau, (S_s)_{s \leq \tau})))$ over all stopping times $\tau \leq T$

under an appropriately chosen measure ${\cal P}$

Classical Solution in discrete Time



$$U_T = A(T, (S_s)_{s \le T}) / (1+r)^T$$

$$U_t = \max\{A(t, (S_s)_{s \le t}) / (1+r)^t, \mathbb{E}^P(U_{t+1} | \mathcal{F}_t)\}$$

for t < T

Stop as soon as the value process reaches the payoff process

Motivation for Multiple Priors Models

Motivation <u>General Framework</u> Exotic Options in Multiple priors Models <u>Conclusions</u> Se

- What is if the market is imperfect?
- Information is imprecise?
- Regulation imposes constraints on trading rules?

Several answers are possible:

- Superhedging
- Utility indifference pricing
- Risk measure pricing

Our approach:



Aim of the paper

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Ambiguity pricing

- Take the perspective of a decision maker who is uncertain about the underlying's dynamics and uses a set of priors instead of a single one
 - Being pessimistic she maximizes the lowest expected return of option

maximize $\inf_{P \in \mathcal{P}} \mathbb{E}^P(A(\tau, S_{\tau})/(1+r)^{\tau})$

- Concentrate on the effect of ambiguity and assume risk neutrality
- Model a consistent market under multiple priors assumption
- Study several exotic options of American style in the framework of ambiguity pricing
- Analyze the difference between classical expected return based pricing and the coherent risk pricing

Results

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Economically

- Ambiguity pricing leads to a valuation under a specific pricing measure
- The pricing measure is rather a part of the solution then of the model itself
- The pricing measure captures the fears of the decision maker and depends on the state and the payoff structure

Mathematically

- The pricing measure might looses the independence property
- Cut off rules are still optimal in this model
- The use of the worst-case measure increases the complexity

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General Framework

Exotic Options in Multiple priors Models

Conclusions

General Framework

The Mathematical Setup

Motivation
General Framework
Exotic Options in Multiple
priors Models

Conclusions

I A probability space $(\Omega, \mathcal{F}, \mathbb{P}_0)$

 $\square \ \Omega = \otimes_{t=1}^{T} \{0, 1\}$ – the set of sequences with values in $\{0, 1\}$

 $\square \mathcal{F}$ – the σ -field generated by all projections $\epsilon_t : \Omega \to \{0, 1\}$

- $\square \mathbb{P}_0$ the uniform on (Ω, \mathcal{F})
- A filtration $(\mathcal{F}_t)_{t=0,...,T}$ generated by the sequence $\epsilon_1, \ldots, \epsilon_t$ with $\mathcal{F}_t = \sigma(\epsilon_1, \ldots, \epsilon_t), \mathcal{F}_0 = \{\emptyset, \Omega\}, \mathcal{F} = \mathcal{F}_T$



Figure 1: Binomial tree

The Mathematical Setup

Motivation
Conorol Framowork
General Flamework
Exotic Options in Multiple priors Models
Conclusions

A convex set of priors ${\mathcal P}$ defined via

 $\mathcal{P} = \left\{ P \in \Delta(\Omega, \mathcal{F}) | P(\epsilon_t = 1 | \mathcal{F}_{t-1}) \in [\underline{p}, \overline{p}] \, \forall t \le T \right\}$

for a fixed interval $[\underline{p},\overline{p}] \subset (0,1)$

- P contains all product measures defined via $P_p(\epsilon_{t+1} = 1 | \mathcal{F}_t) = p$ for a fixed $p \in [p, \overline{p}]$ and all $t \leq T$
- Denote by \overline{P} the measure $P_{\overline{p}}$ and by \underline{P} the measure P_p
- $\epsilon_1, \ldots, \epsilon_t$ are i.i.d under all product measures $P_p \in \mathcal{P}$
 - In general, no independence

Properties of \mathcal{P}

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Lemma 1 The above defined set of priors \mathcal{P} satisfies

1. For all $P \in \mathcal{P}$ $P \sim \mathbb{P}_0$

- All measures in \mathcal{P} agree on the null sets
- We can identify \mathcal{P} with the set of density processes $\mathcal{D} = \{\mathcal{D}_t | t \leq T\}$ where

$$\mathcal{D}_t = \left\{ \left. \frac{dP}{d\mathbb{P}_0} \right|_{\mathcal{F}_t} | P \in \mathcal{P} \right\}$$

■ inf is always a min

Properties of \mathcal{P}

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Lemma 2 \mathcal{P} is time-consistent in the following sense: Let $P, Q \in \mathcal{P}$, $(p_t)_t, (q_t)_t \in (\mathcal{D}_t)_t$. For a fixed stopping time $\tau \leq T$ define the measure R via

$$r_t = \left\{ egin{array}{ccc} p_t & ext{if } t \leq \tau \ rac{p_{ au} q_t}{q_{ au}} & ext{else} \end{array}
ight.$$

Then $R \in \mathcal{P}$.

Time-consistency is equivalent to

a version of *The Law of Iterated Expectations*

fork-stability (FÖLLMER/SCHIED (2004))

rectangularity (EPSTEIN/SCHNEIDER (2003))

 \Rightarrow Allows to change the measure between periods

The Market Structure

Motivation General Framework Exotic Options in Multiple priors Models Conclusions

Ambiguous version of the Cox-Ross-RUBINSTEIN model

A market with 2 assets:

 $\hfill\square$ A riskless asset B with interest rate r>0

 $\hfill\square$ A risky asset S evolving according to $S_0=1$ and

$$S_{t+1} = \begin{cases} S_t \cdot u & \text{if } \epsilon_{t+1} = 1\\ S_t \cdot d & \text{if } \epsilon_{t+1} = 0 \end{cases}$$

Assume $u \cdot d = 1$ and 0 < d < 1 + r < u

 \blacksquare $\overline{P}/\underline{P}$ is the measure with the highest/lowest mean return

Path-dependent increments

Dynamical model adjustment without learning

The Decision Problem

Motivation General Framework Exotic Options in Multiple priors Models

Conclusions

Exercise problem of an ambiguity averse buyer

- For an option paying off $A(t, (S_s)_{s \le t})$ when exercised at t:
 - Choose a stopping time τ^* that maximizes

 $\min_{P \in \mathcal{P}} \mathbb{E}^P(A(\tau, (S_s)_{s \le \tau})/(1+r)^{\tau})$

over all stopping times $\tau \leq T$

Compute

$$U_t^{\mathcal{P}} = \operatorname{esssup}_{\tau \ge t} \operatorname{esssup}_{P \in \mathcal{P}} \mathbb{E}^P(A(\tau, (S_s)_{s \le \tau})/(1+r)^{\tau} | \mathcal{F}_t)$$

– the ambiguity value of the claim at time $t \$

The Solution Method

Motivation General Framework

Exotic Options in Multiple priors Models

Conclusions

Theorem 1 (RIEDEL (2009)) Given a set of measures \mathcal{P} as above and a bounded payoff process X, $X_t = A(t, (S_s)_{s \leq t})/(1+r)^t$, define the **multiple priors Snell envelope** $U^{\mathcal{P}}$ recursively by

$$U_T^{\mathcal{P}} = X_T$$

$$U_t^{\mathcal{P}} = \max\{X_t, \operatorname{essinf}_{P \in \mathcal{P}} \mathbb{E}^P(U_{t+1}^{\mathcal{P}} | \mathcal{F}_t)\} \text{ for } t < T$$
(1)

Then,

1. $U^{\mathcal{P}}$ is the value process of the multiple priors stopping problem for the payoff process *X*, *i.e.*

$$U_t^{\mathcal{P}} = \operatorname{esssup}_{\tau \ge t} \operatorname{essinf}_{P \in \mathcal{P}} \mathbb{E}^P(X_\tau | \mathcal{F}_t)$$

2. An optimal stopping rule is then given by

$$\tau^* = \inf\{t \ge 0 | U_t^{\mathcal{P}} = X_t\}$$

The Solution Method

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Duality result (KARATZAS/ KOU (1998)): There exists a $\hat{P} \in \mathcal{P}$ s.t.

$$U^{\mathcal{P}} = U^{\hat{P}} \quad \mathbb{P}_0 - \text{a.s.}$$

To solve the problem

- Identify the worst-case measure $\hat{P}\in\mathcal{P}$
- Refer to the classical solution

Idea

Identify the worst-case measure for monotone claims

Decompose more complicated claims in monotone parts

Construct the worst-case measure pasting together the worst-case densities of the monotone parts

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General Framework

Exotic Options in Multiple priors Models

Conclusions

Exotic Options in Multiple priors Models

Multiple Expiry Options

Motivation
General Framework
Exotic Options in Multiple priors Models
Conclusions

- Multiple expiry options expiry at some date $\sigma < T$ in the future issuing a new option with conditions specified at $\sigma < T$
- Often used as employee bonus and therefore are subject to trading restrictions
- The value to the buyer/executive differs from the cost to the company of granting the option (HALL/ MURPHY (2002))
- Multiple expiry feature causes a second source of uncertainty:



Dual Expiry Options – Shout Options

Motivation
General Framework
Exotic Options in Multiple
priors Models
Conclusions

- Shout options allow the buyer to shout and freeze the strike at-the-money at any time prior to maturity
- Can be seen as the option to abandon a project to conditions specified by the buyer
- There is uncertainty about the strike at time 0 that is resolved at the time of shouting
- The payoff of the shout option at shouting is an at-the-money put of European style and the problem becomes

maximize
$$A(\sigma, S_{\sigma}) = (S_{\sigma} - S_T)^+ / (1+r)^T$$

over all stopping times $\sigma \leq T$

The task here is rather to start the process optimally than to stop it

Dual Expiry Options – Shout Options

Motivation General Framework Exotic Options in Multiple priors Models Conclusions Since the payoff process is not adapted consider for $t \leq T$

$$X_t = \operatorname{essinf}_{P \in \mathcal{P}} \mathbb{E}^P ((S_t - S_T)^+ / (1 + r)^T | \mathcal{F}_t)$$

= $S_t \cdot g(t, \overline{P})$
= $S_t \cdot \frac{(1 - \overline{p})^T}{(1 + r)^T} \left(\sum_{k=0}^{k(t)} {T - t \choose k} \left(\frac{\overline{p}}{1 - \overline{p}} \right)^k (1 - d^{T-2k}) \right)$

for $k(t) = \left\lfloor \frac{T-t}{2} \right\rfloor$

Lemma 3 For all stopping times $\sigma \leq T$ we have

$$\min_{P \in \mathcal{P}} \mathbb{E}^P(X_{\sigma}) = \min_{P \in \mathcal{P}} \mathbb{E}^P(A(\sigma, S_{\sigma})/(1+r)^T)$$

Dual Expiry Options – Shout Floor

Motivation General Framework Exotic Options in Multiple

Conclusions

priors Models

We can maximize X instead of the original payoff

As a consequence we have

$$U_0^{\mathcal{P}} = \operatorname{essinf}_{P \in \mathcal{P}} \mathbb{E}^P \left(\operatorname{essinf}_{Q \in \mathcal{P}} \mathbb{E}^Q ((S_{\sigma^*} - S_T)^+ | \mathcal{F}_{\sigma^*}) \right)$$
$$= \min_{P \in \mathcal{P}} \mathbb{E}^P \left(S_{\sigma^*} \cdot g(\sigma^*, \overline{P}) \right)$$
$$= \mathbb{E}^{\underline{P}} \left(S_{\sigma^*} \cdot g(\sigma^*, \overline{P}) \right)$$

where σ^{*} is optimal.

The worst-case measure is defined by

$$\hat{P}(\epsilon_{t+1} | \mathcal{F}_t) = \begin{cases} \overline{p} & \text{if } \sigma^* < t \\ \underline{p} & \text{else} \end{cases}$$

Dual Expiry Options – Shout Floor

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Lemma 4 The optimal stopping time for the above problem is given by

$$\sigma^* = \inf\{t \ge 0 : f(t) = x^*\}$$

where

$$f(t) = g(t, \overline{P}) \cdot (\underline{p} \cdot u + (1 - \underline{p}) \cdot d)^t$$

and x^* is the maximum of f on [0,T]

Proof: Generalized parking method and Optional Sampling

Remarks

1. Closed form solutions require exact study of the monotonicity of f

2.
$$1 - \overline{p} \ge (\underline{p} \cdot u + (1 - \underline{p}) \cdot d)$$
 is sufficient to have

$$\sigma^* = 0$$

U–shaped Payoffs

Motivation
General Framework

Exotic Options in Multiple priors Models

Conclusions

- U-shaped payoffs consist of two monotone parts allowing to benefit from change in the underlying independently of the direction of the change
- Often used as speculative instrument before important events



Figure 2: Payoff of Straddle

U–shaped Payoffs

Motivation General Framework Exotic Options in Multiple priors Models

Conclusions

- Up- and Down-movement can increase the value of the claim
- Uncertainty does not vanish over time

Lemma 5 The value process is Markovian. For every $t \leq T$ the value function $v(t, \cdot)$ is quasi-convex and there exists a sequence $(\hat{x}_t)_{t \leq T}$ s.t. $v(t, \cdot)$ increases on $\{x_t > \hat{x}_t\}$ and decreases else.

Proof: Backward induction

Proof uses explicitly the binomial structure of the model

As a consequence we obtain

$$\hat{P}(\epsilon_{t+1} = 1 | \mathcal{F}_t) = \begin{cases} \overline{p} & \text{if } S_t < \hat{x}_t \\ \underline{p} & \text{if } S_t \ge \hat{x}_t \end{cases}$$

U–shaped Payoffs

WOUVALION
General Framework
Exotic Options in Multiple
phors models

Conclusions

Mativation

- The worst-case measure is mean-reverting (in a wider sense)
- The drift changes every time S hits a barrier and can happen arbitrary often
- Fears of the decision maker are opposite to the market movement
- Increments are not independent anymore
- Idea: Use the generalized parking method again or upper and lower bounds

Motivation

General Framework

Exotic Options in Multiple priors Models

Conclusions

Conclusions

Conclusions

General Framework Exotic Options in Multiple priors Models

Conclusions

Motivation

Conclusions

- A model of a multiple priors market provided
- A method to evaluate options in imperfect markets proposed
- Pricing measure for several classes of payoffs derived
- Worst-case measure is path-dependent in general
- The structure of the stopping times carries over in this model
- This is, however, due to the model and not a general result

Future work

Motivation	_
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General Framework	_
Exotic Options in Multiple	
priors Models	_
	_
Conclusions	
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- Continuous-time analysis Brownian motion setting
- Infinite time modeling in continuous time
 - □ Allows for closed form solutions and comparative statics
 - □ Mathematical traps due to multiple measure structure
 - More modeling necessary to build a meaningful mathematical model

Motivation
General Framework
Exotic Options in Multiple priors Models
Conclusions

Thank you for your attention!