

15.094J: Robust Modeling, Optimization, Computation

Lecture 16: Robust Option Pricing

Outline

- 1 What are options?
- 2 The classical theory of options pricing
- 3 The Idea of no-arbitrage
- 4 Optimal Replication
- 5 Robust Options Pricing
- 6 Extensions
- 7 Discussion
 - Computational Tractability
 - Modeling Flexibility
- 8 Computational Results

What are options and why they are important?

- A call option on a stock of strike \$50 with maturity 3 months, gives the right to buy the stock 3 months from now at \$50. So, if the price at that time is \$90, then there is a profit of \$40.
- A put option gives the right to sell the stock.
- American versus European options.
- Call options are a widespread method of compensation for executives.
- Put options provide insurance.
- The derivatives industry: 10 trillion dollar industry

Basics

- An option is a contract defined on a set of *predetermined underlying securities*

$$\mathbf{S} = \{S_i\}_{i=1,\dots,M},$$

and is associated with a *payoff function*

$$P_f(\mathbf{S}, \mathbf{K}, T),$$

where \mathbf{K} and T are a set of parameters specified in the contract.

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Example. European Call Option

The option holder has the right to buy a unit of stock at a price of K , at time T , even when the market price is S_T . The payoff, then, is

$$P_f(S, K, T) = \max(S_T - K, 0).$$

Dynamics

- Given a stock of price S_t

$$S_{t+1} = \begin{cases} u \cdot S_t & \text{with probability } q \\ d \cdot S_t & \text{with probability } 1 - q \end{cases}$$

- Bond with price \$1

$$\$1 \rightarrow r, \quad r \text{ Riskless Return}$$

Payoff of an option

Consider an option (also called derivative security):

$$C_{t+1} = \begin{cases} C_u(S_t) & \text{with probability } q \\ C_d(S_t) & \text{with probability } 1 - q \end{cases}$$

- What should the price of the option be?
- What should it depend on?

European call option

- $f(S) = \begin{cases} 0 & S \leq K \\ S - K & S > K \end{cases}$
- $C_u = C_u(S) = \max(u \cdot S - K, 0)$
- $C_d = C_d(S) = \max(d \cdot S - K, 0)$.

The Idea of no-arbitrage

- The story of the Princeton professor....
- Consider a portfolio that has x \$ worth of the stock and B \$ worth of the bond that pay off $B \cdot r$.
- Cost of portfolio: $x + B$
- Return of portfolio:

$$\begin{cases} x \cdot u + B \cdot r & \text{with probability } q \\ x \cdot d + B \cdot r & \text{with probability } 1 - q \end{cases}$$

Critical Idea

- Choose x, B so that we create the same payoff structure as the option, i.e.,

$$x \cdot u + B \cdot r = C_u$$

$$x \cdot d + B \cdot r = C_d$$

- Solving the Linear System

$$x^* = \frac{C_u - C_d}{u - d}, \quad B^* = \frac{u \cdot C_d - d \cdot C_u}{(u - d)r}$$

Critical Idea, continued

- What should the price of the option be?
- By No-Arbitrage

$$\underbrace{C}_{\text{Cost of Option}} = \underbrace{x^* + B^*}_{\text{Cost of Portfolio}}$$

- Since they have identical payoffs, they should have the same cost; otherwise there exists an arbitrage opportunity.

The Price

- By no arbitrage:

$$\begin{aligned}
 C &= \frac{C_u - C_d}{u-d} + \frac{u \cdot C_d - d \cdot C_u}{(u-d)r} \\
 &= \frac{1}{r} [p \cdot C_u + (1-p)C_d] \quad , \quad p = \frac{r-d}{u-d}
 \end{aligned}$$

- Price of option is the present value of the expected payoff of the option, but not using the original probability q , but probability p , the risk neutral measure.

Multiple periods

- Consider now T periods
- Payoff of European call option

$$f(S_T) = (S_T - K)^+ = \max(S_T - K, 0)$$

- Price

$$C = \frac{1}{r^T} \sum_{n=0}^T \binom{T}{n} p^n (1-p)^{T-n} (u^n d^{T-n} S - K)^+$$

- Independent of probability q

Limitations

- Options depending on many securities problematic computationally.
- Modeling transaction costs, liquidity issues and market restrictions problematic conceptually and computationally.

The Model for stock returns

- Consider discrete time $\{0, 1, 2, \dots, T\}$.
- Let \tilde{r}_t^S be the return at t ; i.e., the return from period $[t, t + 1)$.
- Assuming $\{\tilde{r}_1^S, \tilde{r}_2^S, \dots, \tilde{r}_\tau^S\}$ are independent, we have from the central limit theorem,

$$\frac{\sum_{i=1}^{\tau} \log(1 + \tilde{r}_i^S) - \tau \cdot \mu_{\log}}{\sigma_{\log} \cdot \sqrt{\tau}} \sim N(0, 1),$$

where μ_{\log} , σ_{\log} are mean and standard deviation of $\log(1 + \tilde{r}_i^S)$, respectively.

The Model for stock returns, continued

- The CLT motivates us to consider constraints of the form

$$\left| \frac{\log \tilde{R}_\tau^S - \tau \cdot \mu_{\log}}{\sigma_{\log} \cdot \sqrt{\tau}} \right| \leq \Gamma_\tau \quad \forall \tau,$$

- $\tilde{R}_\tau^S = \prod_{i=1}^\tau (1 + \tilde{r}_i^S)$, is the cumulative return up to time τ and Γ_τ is some parameter.
- Other constraints can be based on the the assumption of a bounded support

$$\mu_r - \Gamma_\tau \sigma_r \leq \frac{\tilde{R}_\tau^S}{\tilde{R}_{\tau-1}^S} \leq \mu_r + \Gamma_\tau \sigma_r \quad \forall \tau.$$

The Model for stock returns, continued

- Uncertainty set for stock returns:

$$\mathbb{U}^1 = \left\{ \tilde{R}_t^S \left| \begin{array}{ll} \underline{R}_t^S \leq \tilde{R}_t^S \leq \overline{R}_t^S, & \forall t = 1 \dots T \\ \underline{r}_t^S \cdot \tilde{R} \leq \tilde{R}_t^S \leq \overline{r}_t^S \cdot \tilde{R}_{t-1}^S, & \forall t = 1 \dots T \\ \underline{R}_{t,\tau}^S \cdot \tilde{R}_\tau \leq \tilde{R}_t \leq \overline{R}_{t,\tau}^S \cdot \tilde{R}_\tau, & \forall \{(t, \tau) | \tau < t, t = 1 \dots T\} \end{array} \right. \right\}$$

where $\underline{R}_t^S = e^{t \cdot \mu_{\log} - \Gamma \cdot \sqrt{t} \cdot \sigma_{\log}}$, $\overline{R}_t^S = e^{t \cdot \mu_{\log} + \Gamma \cdot \sqrt{t} \cdot \sigma_{\log}}$, $\underline{r}_t^S = \mu_r - \Gamma_t \cdot \sigma_r$,

$\overline{r}_t^S = \mu_r + \Gamma_t \cdot \sigma_r$,

$R_{t,\tau}^S = (t - \tau) \cdot \mu_r - \Gamma_t \cdot \sigma_r \cdot \sqrt{t - \tau}$ and

$\overline{R}_{t,\tau}^S = (t - \tau) \cdot \mu_r + \Gamma_t \cdot \sigma_r \cdot \sqrt{t - \tau}$.

The idea of ϵ -arbitrage

- Replicating portfolios and incomplete markets
 - Exact replication may not be possible.
- The idea of ϵ -arbitrage.
 - Compute the best possible replicating portfolio, when the stock returns lie in an uncertainty set.
 - The resulting replication error stands for the ϵ -arbitrage.
- ϵ can be seen as a measure of incompleteness of the market.

The Problem of Optimal Replication

- Given $P(S_T, K)$ is the payoff of the option.
- Define x_t^S and x_t^B are the amounts invested in the stock and the bond during the period $[t, t+1)$.
- W_T is the value of the replicating portfolio.

$$\begin{aligned}
 & \min_{\{x_t^S, x_t^B, y_t\}} \max_{\{\tilde{R}_t^S \in \mathbb{U}^1\}} |P(S_T, K) - W_T| \\
 & \text{s.t.} \quad W_T = x_T^S + x_T^B \\
 & \quad x_t^S = (1 + \tilde{r}_{t-1}^S) (x_{t-1}^S + y_{t-1}), \forall t = 1, \dots, T, \\
 & \quad x_t^B = (1 + r_{t-1}^B) (x_{t-1}^B - y_{t-1}), \forall t = 1, \dots, T,
 \end{aligned}$$

Robust Options Pricing

- European Call option: $P(\tilde{S}, K) = (\tilde{S}_T - K)^+$.
- The optimization problem

$$\min_X \max_{U^1} \left| (\tilde{S}_T - K)^+ - W_T \right|$$

$$\text{s.t.} \quad W_T = x_T^S + x_T^B,$$

$$x_t^S = (1 + \tilde{r}_{t-1}^S) (x_{t-1}^S + y_{t-1}), \forall t = 1 \dots T,$$

$$x_t^B = (1 + r_{t-1}^B) (x_{t-1}^B - y_{t-1}), \forall t = 1 \dots T.$$

Robust Options Pricing, continued

- Variable transformations :

$$\alpha_t^S = \frac{x_t^S}{R_t^S}, \alpha_t^B = \frac{x_t^B}{R_t^B}, \beta_t = \frac{y_t}{R_t^S}, \text{ where } \tilde{R}_t^S = \prod_{i=0}^{t-1} (1 + \tilde{r}_i^S), \text{ and } R_t^B = \prod_{i=0}^{t-1} (1 + r_i^B).$$

- After substitution:

$$\begin{aligned} & \min_{\{\alpha_t^S, \alpha_t^B, \beta_t\}} \max_{\{\tilde{R}_t^S \in \mathbb{U}^1\}} \left| \left(S_0 \tilde{R}_T^S - K \right)^+ - \left(\tilde{R}_T^S \alpha_T^S + R_T^B \alpha_T^B \right) \right| \\ & \text{s.t.} \quad \alpha_t^S = \alpha_{t-1}^S + \beta_{t-1}, \quad \forall t = 1, \dots, T, \\ & \quad \alpha_t^B = \alpha_{t-1}^B - \beta_{t-1} \frac{\tilde{R}_{t-1}^S}{R_{t-1}^B}, \quad \forall t = 1, \dots, T. \end{aligned}$$

Robust Options Pricing, continued

- Substituting all intermediate α_t^B , α_t^S :

$$\min_{\{\alpha_t^S, \alpha_t^B, \beta_t\}} \max_{\{\tilde{R}_t^S \in \mathbb{U}^1\}} \left| \left(S_0 \tilde{R}_T^S - K \right)^+ - \left(\alpha_0^S + \sum_{t=1}^T \beta_{t-1} \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=0}^{T-1} \beta_t \frac{R_T^B}{R_t^B} \tilde{R}_t^S \right|.$$

- Inner problem:

$$\begin{aligned} \min \quad & \kappa \\ \text{s.t.} \quad & \kappa \geq \left(S_0 \tilde{R}_T^S - K \right)^+ - \left(\alpha_0^S + \sum_{t=1}^T \beta_{t-1} \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=0}^{T-1} \beta_t \frac{R_T^B}{R_t^B} \tilde{R}_t^S, \quad \forall \tilde{R}_t^S \in \mathbb{U}^1, \\ & \kappa \leq - \left(\left(S_0 \tilde{R}_T^S - K \right)^+ - \left(\alpha_0^S + \sum_{t=1}^T \beta_{t-1} \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=0}^{T-1} \beta_t \frac{R_T^B}{R_t^B} \tilde{R}_t^S \right), \quad \forall \tilde{R}_t^S \in \mathbb{U}^1. \end{aligned}$$

Robust Options Pricing, continued

- Model the piecewise-linear function $\left(S_0 \tilde{R}_T^S - K\right)^+$,
- We partition the uncertainty set \mathbb{U}^1

$$\mathbb{U}_a^1 = \mathbb{U}^1 \cap \left\{ \tilde{R}_T^S \geq \frac{K}{S_0} \right\}, \quad \mathbb{U}_b^1 = \mathbb{U}^1 \cap \left\{ \tilde{R}_T^S \leq \frac{K}{S_0} \right\}.$$

Robust Options Pricing, continued

- Using this partition, we obtain the following equivalent formulation

$$\min_{\{\alpha_0^S, \alpha_0^B, \beta_t\}} \epsilon$$

s. t.

$$\epsilon \geq \left(S_0 \tilde{R}_T^S - K \right) - \left(\alpha_0^S + \sum_{t=0}^{T-1} \beta_t \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=0}^{T-1} \beta_t \frac{R_T^B}{R_t^B} \tilde{R}_t^S, \quad \forall \tilde{R}_t^S \in \mathbb{U}_a^1,$$

$$\epsilon \geq - \left(\left(S_0 \tilde{R}_T^S - K \right) - \left(\alpha_0^S + \sum_{t=0}^{T-1} \beta_t \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=0}^{T-1} \beta_t \frac{R_T^B}{R_t^B} \tilde{R}_t^S \right), \quad \forall \tilde{R}_t^S \in \mathbb{U}_a^1,$$

$$\epsilon \geq - \left(\alpha_0^S + \sum_{t=0}^{T-1} \beta_t \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=0}^{T-1} \beta_t \frac{R_T^B}{R_t^B} \tilde{R}_t^S, \quad \forall \tilde{R}_t^S \in \mathbb{U}_b^1,$$

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Robust Options Pricing, continued

- Can be converted to an equivalent linear optimization problem.
- Resulting size: $16T + 4$ decision variables and $4T + 4$ constraints.

Pricing Asian options

- Asian Call option: $P(\tilde{S}, K) = (S_0 \tilde{R}_{\text{ave}}^S - K)^+$, where

$$\tilde{R}_{\text{ave}}^S = \sum_{t=1}^T \frac{\tilde{R}_t^S}{T}.$$

- Again leads to a linear formulation with size that scales linearly in T .

Pricing Lookback options

- $P(\tilde{S}, K) = (S_0 \tilde{R}_{\max}^S - K)^+$, $\tilde{R}_{\max}^S = \max_{t=1 \dots T} \{\tilde{R}_t^S\}$.
- We obtain the following formulation

$$\min_{\{\alpha_0^S, \alpha_0^B, \beta_t\}} \epsilon$$

s.t.

$$\forall k = 1 \dots T$$

$$\epsilon \geq (S_0 \tilde{R}_k^S - K) - \alpha_0^S \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=1}^T \beta_{t-1} \left(\frac{R_t^B}{R_t^B} \tilde{R}_t^S - \tilde{R}_T^S \right) \quad \forall \mathbb{U}_k^1 \cap \left\{ \tilde{R}_k^S \geq \frac{K}{S_0} \right\}$$

$$\epsilon \geq - (S_0 \tilde{R}_k^S - K) + \alpha_0^S \tilde{R}_T^S + \alpha_0^B R_T^B - \sum_{t=1}^T \beta_{t-1} \left(\frac{R_t^B}{R_t^B} \tilde{R}_t^S - \tilde{R}_T^S \right) \quad \forall \mathbb{U}_k^1 \cap \left\{ \tilde{R}_k^S \geq \frac{K}{S_0} \right\}$$

$$\epsilon \geq - \left(\alpha_0^S + \sum_{t=1}^T \beta_{t-1} \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=1}^T \beta_{t-1} \frac{R_t^B}{R_t^B} \tilde{R}_t^S \quad \forall \mathbb{U}_k^1 \cap \left\{ \tilde{R}_k^S \leq \frac{K}{S_0} \right\}$$

$$\epsilon \geq - \left(- \left(\alpha_0^S + \sum_{t=1}^T \beta_{t-1} \right) \tilde{R}_T^S - \alpha_0^B R_T^B + \sum_{t=1}^T \beta_{t-1} \frac{R_t^B}{R_t^B} \tilde{R}_t^S \right) \quad \forall \mathbb{U}_k^1 \cap \left\{ \tilde{R}_k^S \leq \frac{K}{S_0} \right\}$$

Pricing American Put options

- Can exercise the option at any time up to the time of exercise T .
- The payoff is then given by $P(\tilde{S}, K) = (K - \tilde{S}_\tau)$, where τ is the time of exercise.
 - τ depends on the utility of the option holder.
- In the absence of any information about the utility function of the option holder, we seek to find a replicating portfolio for all possible exercise policies.

Pricing American Put options

- Proceeding as before, we obtain the following formulation

$$\min_{\{\alpha_t^S, \alpha_t^B, \beta_t\}} \max_{\{\tilde{R}_t^S\} \in \mathcal{U}^1} \max_{\tau=1, \dots, T} \left| \left(K - S_0 \tilde{R}_\tau \right)^+ - \left(\tilde{R}_\tau^S \alpha_\tau^S + R_\tau^B \alpha_\tau^B \right) \right|$$

$$\text{s.t.} \quad \alpha_t^S = \alpha_{t-1}^S + \beta_{t-1}, \quad \forall t = 1, \dots, T,$$

$$\alpha_t^B = \alpha_{t-1}^B - \beta_{t-1} \frac{\tilde{R}_t^S}{R_t^B}, \quad \forall t = 1, \dots, T,$$

where τ is the time of exercise.

- The size of the resulting formulation will be quadratic in T .

Pricing Barrier options

- These options become inactive as soon as a condition(\mathbb{C}) of the form $S_t \leq a$ or $S_t \geq b$ is reached.
- The problem of optimal replication then reduces to

$$\min_{\{x_t^S, x_t^B, y_t\}} \max_{\{\tilde{R}_t^S \in \mathbb{U}^1 \cap \mathbb{C}\}} |P(S_T, K) - W_T|$$

because if \mathbb{C} is not satisfied the option ceases to exist and one need not worry about replicating its payoff.

- We obtain a linear formulation that scales linearly in T .

Pricing Multidimensional options

- Pricing options that depend on M underlying assets is difficult to price, for large M , using current methods because of
 - unavailability of an analytic solution and
 - the curse-of-dimensionality which prevents one from using dynamic programming.
- Proceeding as before, we seek to obtain the optimal solution of the following optimization problem

$$\begin{aligned}
 & \min_{\{x_t^m, y_t^m\}} \quad \max_{\{\tilde{r}_t^m\} \in \mathbb{U}^M} |P_f(\{S^i\}_{i=1, \dots, M}, K) - W_T| \\
 & \text{s.t.} \quad W_T = \sum_{m=0}^M x_T^m, \\
 & \quad x_t^m = (1 + \tilde{r}_{t-1}^m) \cdot (x_{t-1}^m + y_{t-1}^m) \quad \forall t = 1, \dots, T, \quad \forall m = 1, \dots, M, \\
 & \quad x_t^0 = (1 + \tilde{r}_{t-1}^0) \cdot \left(x_{t-1}^0 - \sum_{m=1}^M y_{t-1}^m \right) \quad \forall t = 1, \dots, T.
 \end{aligned}$$

Multidimensional options

- \mathbf{C} : the covariance matrix of the single period returns.
- Uncertainty set \mathbb{U}^M

$$\mathbb{U}^M = \left\{ \tilde{\mathbf{R}}_t \left| \begin{array}{ll} \|\mathbf{C}(\tilde{\mathbf{R}}_1 - \bar{\mathbf{R}}_1)\| \leq \Gamma, & \forall t = 2, \dots, T, \forall m = 1, \dots, M, \\ \underline{r}_t^m \tilde{R}_{t-1}^m \leq \tilde{R}_t^m \leq \overline{r}_t^m \tilde{R}_{t-1}^m & \forall t = 2, \dots, T, \forall m = 1, \dots, M, \\ \underline{R}_t^m \leq \tilde{R}_t^m \leq \overline{R}_t^m, & \forall t = 1, \dots, T, \forall m = 1, \dots, M. \end{array} \right. \right\}.$$

- By choosing the ℓ_1 or ℓ_∞ we obtain LO formulations.

Computational Complexity

- Methodology scales polynomially with the dimension of the option and the discretization, unlike DP.

Option Type	European	Asian	Lookback	American	Index	American Index
Size	$O(T)$	$O(T)$	$O(T^2)$	$O(T^2)$	$O(M \cdot T)$	$O(M \cdot T^2)$

T : the number of time periods the option is written for.

M : the number of different assets required to define the option.

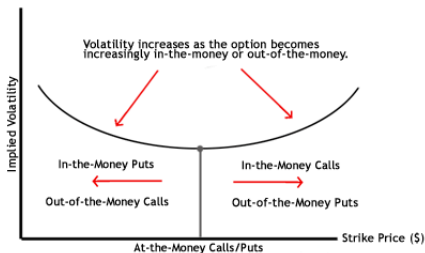
Size : Number of variables and constraints.

Modeling Flexibility

- Modeling many types of options - Barrier, Lookback etc.
- Modeling implied volatility smile
- Modeling transaction costs and other market restrictions, still obtaining LO formulations.

Modeling Implied Volatility smile

- *Implied volatility* of an option is simply that volatility that makes the Black-Scholes (BS) model price exactly equal to the observed market price.
 - Expect flat lines, if market participants use BS Model.
- Plotted across strike prices, they exhibit *smiles* or *smirks*.



Modeling the implied volatility smile

- Risk Aversion as an explanation for the Implied Volatility Smile.
- Lower value for Γ implies that the user is willing to take higher risk by ignoring the variability of stock prices.
- Higher value of Γ indicates that the user seeks a price that will allow him to replicate the payoff of the option for a larger range of stock prices.
- Historical evidence.
 - No smile was observed before the crash of 1987.
 - After the crash, the smile started appearing.

Modeling the implied volatility smile, continued

Risk Aversion as an explanation for the Implied Volatility Smile

- Empirical results indicate that a quadratic dependence of $\Gamma_{implied}$ with $\frac{K}{S_0}$ would be adequate to characterize the risk aversion of investor towards different strike prices.

- The following function is used to describe the relationship:

$$\Gamma(K) = \theta_0 + \theta_1 \frac{K - S_0}{S_0} + \theta_2 \left(\frac{K - S_0}{S_0} \right)^2, \quad \theta_2 \geq 0.$$

- The quantity $\frac{K - S_0}{S_0}$ captures the distance between the strike and the spot price and is also called as *moneyness* in the literature.

Experiments setup

- We perform the following experiments:
 - Experiment 1 : Compare with actual market prices for European call options.
 - Experiment 2 : Compare with actual market prices for American put options.

Experiments setup

- All the experiments have a training stage and a testing stage.
 - In the training stage, we choose a random set of strike prices and calibrate (compute $\theta_0, \theta_1, \theta_2$) our model to it.
 - In the testing stage, we use our model to price the options for the remaining strikes.

Experiment 1

- The underlying security is Microsoft stock.
- The number of periods $T = 18$ weeks.
- The initial price of underlying security $S_0 = 21.4$.
- Strike price of options K : ranges from 2.5 to 30.

Experiment 1

Out of Sample

No.	T	K/S	$\Gamma_{implied}$	Mkt Price	Model Price	Error
1	18	0.654	2.45	7.475	7.48	0.005
2	18	0.794	2.056	4.8	4.797	-0.003
3	18	0.888	1.75	3.25	3.232	-0.018
4	18	0.981	1.66	1.97	1.968	-0.002
5	18	1.028	1.65	1.47	1.462	-0.008
6	18	1.121	1.73	0.735	0.749	0.014

In Sample

No.	T	K/S	$\Gamma_{implied}$	Mkt Price	Model Price	Error
1	18	0.607	2.77	8.425	8.42	-0.005
2	18	0.701	2.25	6.55	6.561	0.011
3	18	0.841	1.921	4	3.984	-0.016
4	18	0.935	1.69	2.56	2.556	-0.004
5	18	1.285	2.19	0.155	0.152	-0.003

Experiment 2

- The underlying security is MSFT stock.
- The number of periods $T = 25$ weeks.
- The initial price of underlying security $S_0 = 24.8$.
- Strike price of options K : ranges from 7.5 to 50.

Experiment 2

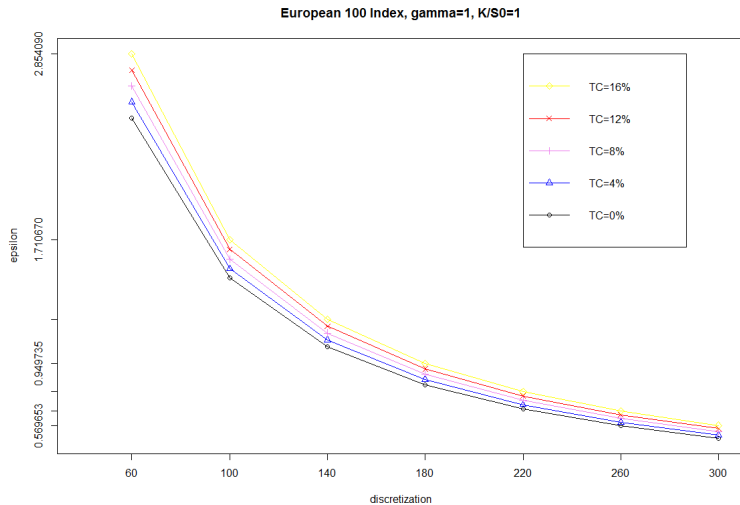
Out of Sample

No.	T	K/S	$\Gamma_{implied}$	Mkt Price	Model Price	Error
1	25	0.605	2.62	0.17	0.201	0.031
2	25	0.806	1.83	0.695	0.589	-0.106
3	25	0.968	1.6	1.895	1.764	-0.132
4	25	1.008	1.59	2.365	2.266	-0.099
5	25	1.21	1.9	5.85	5.939	0.089
6	25	1.411	2.87	10.5	10.703	0.203
7	25	1.815	7.7	20.45	20.303	-0.147

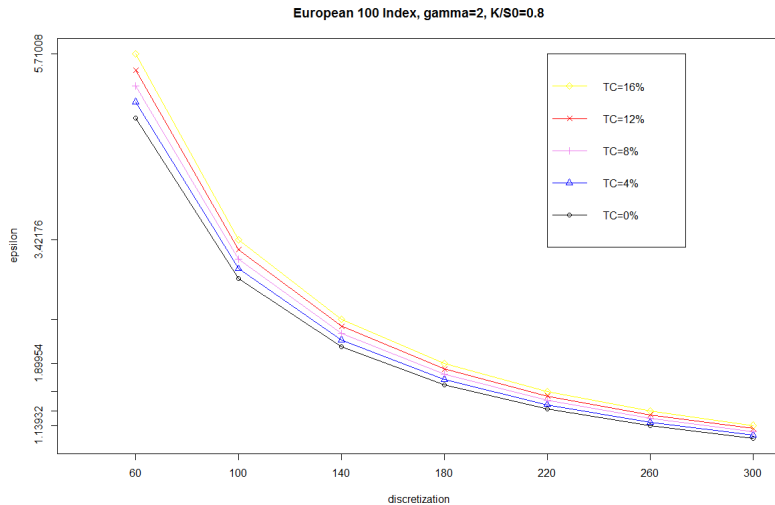
In Sample

No.	T	K/S	$\Gamma_{implied}$	Mkt Price	Model Price	Error
1	25	0.504	3.24	0.065	0.17	0.105
2	25	0.706	2.15	0.34	0.305	-0.035
3	25	0.766	1.94	0.525	0.442	-0.083
4	25	0.847	1.74	0.905	0.778	-0.127

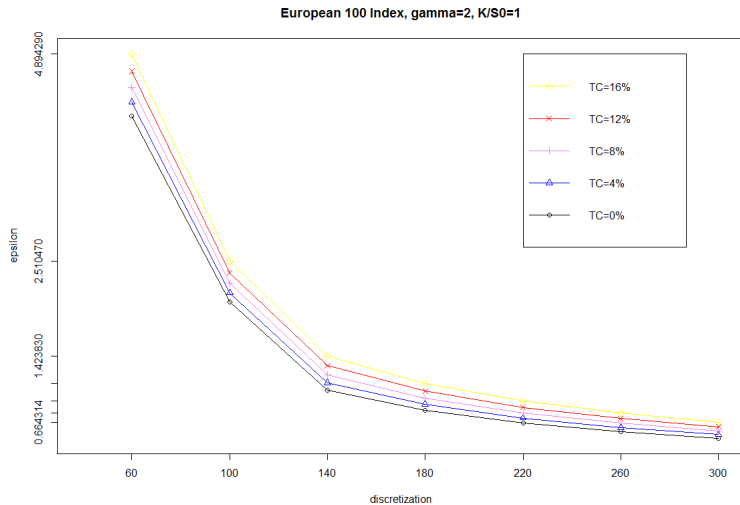
Dependence on Discretization and Transaction Costs



Dependence on Discretization and Transaction Costs



Dependence on Discretization and Transaction Costs



Conclusions

- Tractable approach to price options while accounting for the risk attitudes of the option writer.
- Approach scales polynomially (as opposed to exponentially) with the dimension of the original pricing problem.
- We provide a potential explanation for the phenomenon of the implied volatility smile, and support experimental evidence for the same.