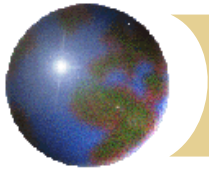


Machine Learning in Business

John C. Hull

Chapter 10

Applications in Finance

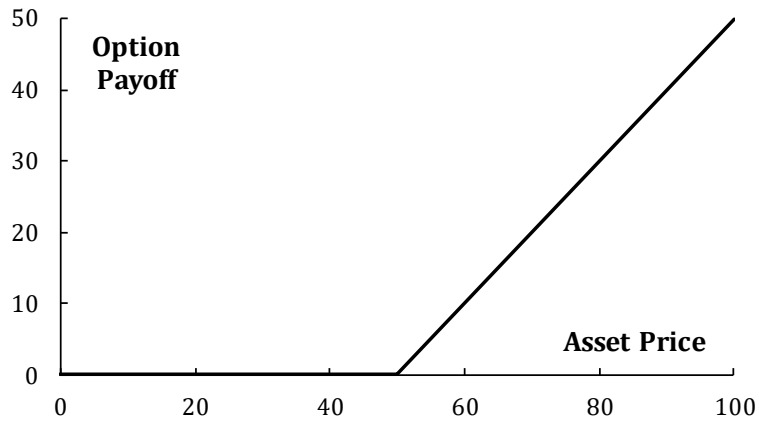


Call and Put Options

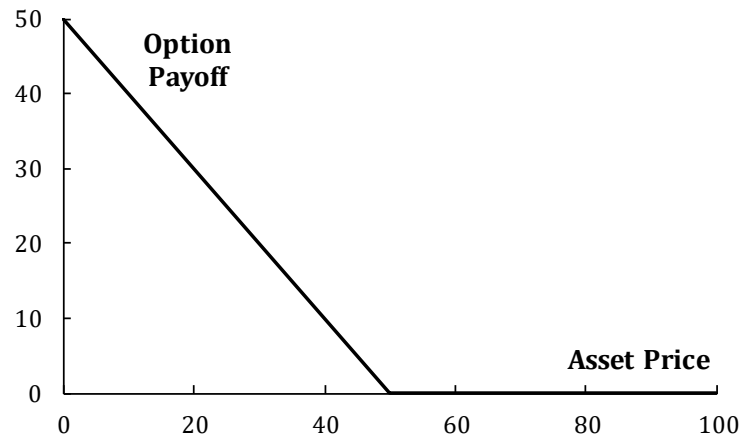
- ✚ Call option is an option to buy an asset on a certain future date (the maturity) for a certain price (the strike price)
- ✚ Put option is an option to sell an asset on a certain future date (the maturity) for a certain price (the strike price)



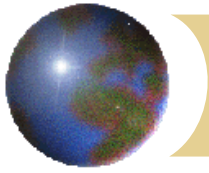
The Payoffs



Call Option

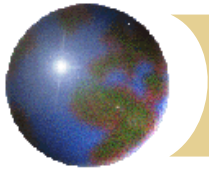


Put Option



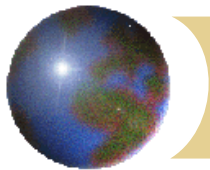
Importance of volatility

- ⊕ The asymmetry in the payoff means that volatility, σ , is important in determining option prices
- ⊕ As volatility increases, the price of a call or put option increases



Moneyiness

- ⊕ Moneyiness is a measure of the extent to which an option is likely to be exercised
- ⊕ Popular definitions (S = asset price, K = strike price)
 - ⊠ At-the money: $S = K$
 - ⊠ In-the money: $S > K$ for call options and $S < K$ for put options
 - ⊠ Out-of-the-money: $S < K$ for call options and $S > K$ for put options



Delta

- ✚ The delta of a portfolio is the sensitivity of the portfolio to the underlying asset price
- ✚ A delta-neutral portfolio is not sensitive to small changes in the underlying asset price
- ✚ If

$$d_1 = \frac{\ln(S/K) + (r - q + \sigma^2/2)T}{\sigma\sqrt{T}}$$

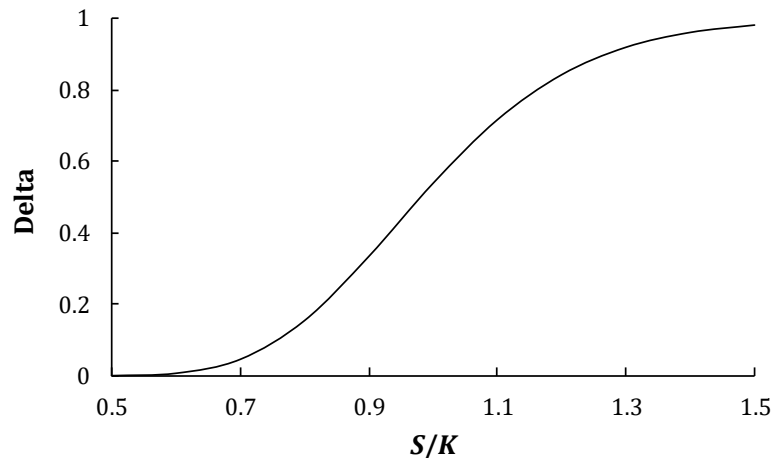
$$\text{Delta (call)} = e^{-qT} N(d_1)$$

$$\text{Delta (put)} = e^{-qT} [N(d_1) - 1]$$

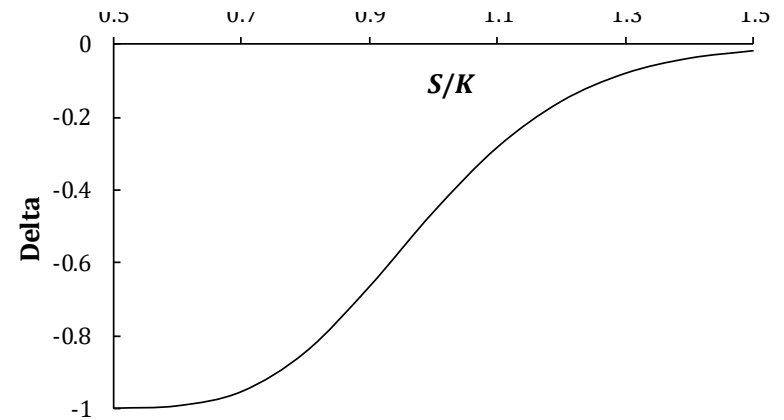
where T is time to maturity, r is risk-free rate, and q is dividend yield



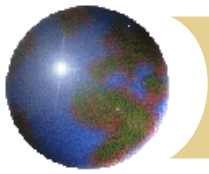
Delta continued



Call Option

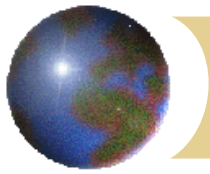


Put Option



Delta continued

- ✚ Traders can make themselves delta-neutral by trading the underlying asset
- ✚ They also use delta as a measure of moneyness for call and put options
 - ✚ Call is at-the-money when $\text{delta} = 0.5$
 - ✚ Call is in-the-money when $\text{delta} > 0.5$
 - ✚ Call is out-of-the-money when $\text{delta} < 0.5$
 - ✚ Put is at-the-money when $\text{delta} = -0.5$
 - ✚ Put is in-the-money when $\text{delta} < -0.5$
 - ✚ Put is out-of-the-money when $\text{delta} > -0.5$



Implied volatilities

- ✚ The Black-Scholes-Merton price of an option depends on S , K , r , q , σ , and T
- ✚ S , K , r , and T are known for an option
- ✚ q can be estimated from futures or forward contracts
- ✚ This means that σ is the only unknown.
- ✚ There is a one-to-one correspondence between option price and σ
- ✚ The implied volatility of an option is the volatility that when substituted into Black-Scholes-Merton formula gives the price of the option in the market

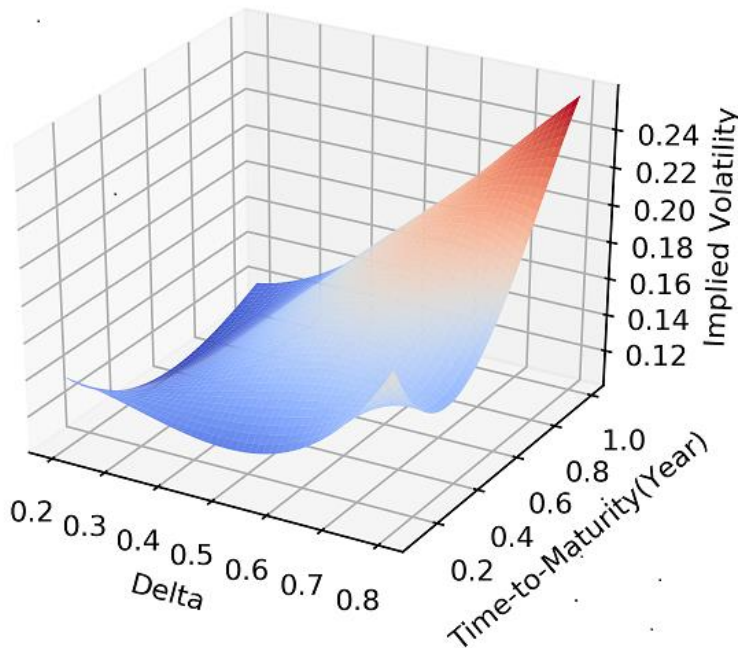


Implied volatilities continued

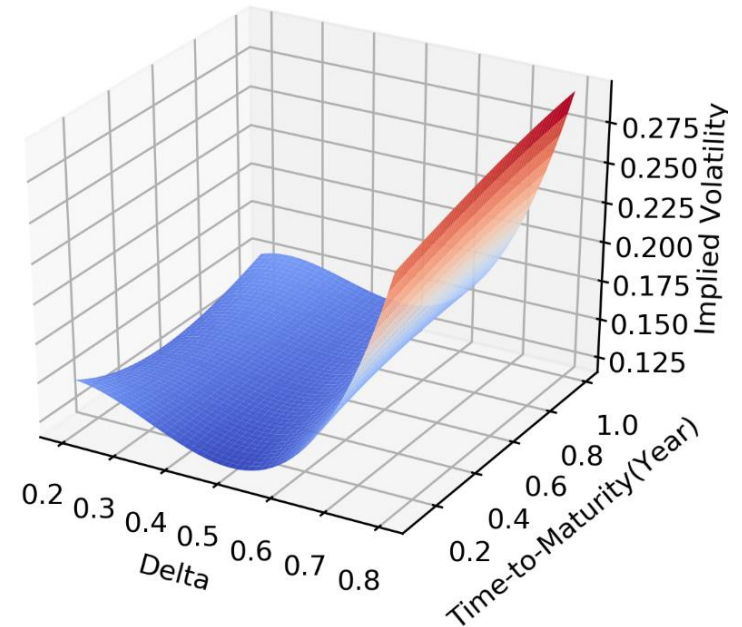
- ⊕ If Black-Scholes-Merton was used for pricing, all options on an asset would have the same implied volatility
- ⊕ In fact, there is quite a variation in implied volatilities
- ⊕ Nevertheless implied volatilities are used to communicate prices and it is therefore important for traders to monitor implied volatilities
- ⊕ The volatility surface shows implied volatilities as a function of
 - ⊠ Moneyness (measured by delta)
 - ⊠ Time to maturity, T



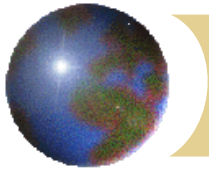
Volatility Surfaces for S&P 500



Jan 31, 2019



June 25, 2019



Volatility Surface Movements

- ⊕ When the price of the underlying asset increases (decreases), implied volatilities tend to decrease (increase)
- ⊕ However not all implied volatilities change by the same amount
- ⊕ This explains the variation in volatility surfaces



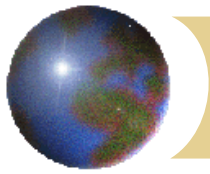
Understanding Volatility Surface Movements

- ⊕ To understand volatility surface movements we used data on S&P 500 call options to construct a neural network
- ⊕ Input layer:
 - ⊠ Daily asset price return
 - ⊠ Moneyness (measured by delta)
 - ⊠ Time to maturity
- ⊕ Output layer:
 - ⊠ Change in implied volatility



Details

- ✚ 3 hidden layers
- ✚ 20 neurons per layer
- ✚ Observations from 2014-2019
- ✚ Randomly sampled 100 options per day
- ✚ 125,700 options in total
- ✚ 60% for training set
- ✚ 20% for validation set
- ✚ 20% for test set
- ✚ Z-score scaling

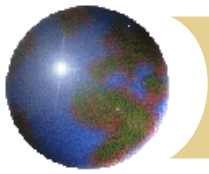


Sample of raw data

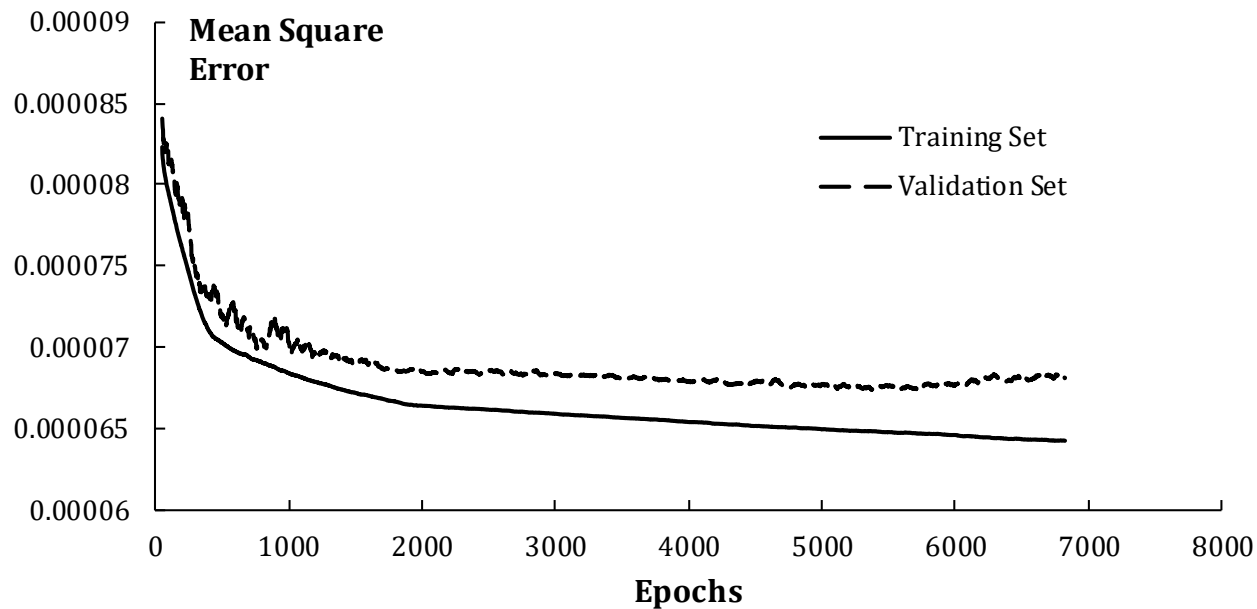
<i>Date</i>	<i>Return (%) on S&P 500</i>	<i>Maturity (years)</i>	<i>Delta</i>	<i>Change in implied volatility (bps)</i>
Jan 28, 2015	0.95	0.608	0.198	-31.1
Sept 8, 2017	1.08	0.080	0.752	-54.9
Jan 24, 2018	0.06	0.956	0.580	-1.6
Jun 24, 2019	-0.95	2.896	0.828	40.1

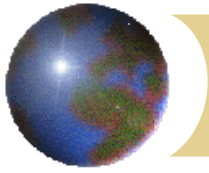
After scaling:

<i>Date</i>	<i>Return (%) on S&P 500</i>	<i>Maturity (years)</i>	<i>Delta</i>	<i>Change in implied volatility (bps)</i>
Jan 28, 2015	1.091	-0.102	-1.539	-31.1
Sept 8, 2017	1.247	-0.695	0.445	-54.9
Jan 24, 2018	0.027	0.289	-0.171	-1.6
Jun 24, 2019	-1.176	2.468	0.716	40.1



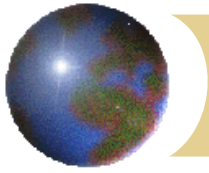
Mean Squared Error (Training stopped after 5,826 epochs)





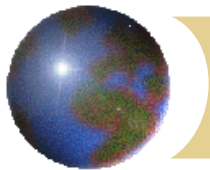
Results

- ⊕ Test set were a modest 14% improvement over a simple analytic model proposed by Hull and White in 2017
- ⊕ However, when the VIX index on Day t was used as a feature to predict changes between Day t and Day $t+1$ there was a considerable improvement
- ⊕ The behavior of the volatility surface is different in high and low volatility environments



Second Application: Hedging

- ⊕ Suppose we do not know how to calculate delta
- ⊕ Can reinforcement hedging find a good hedging strategy?

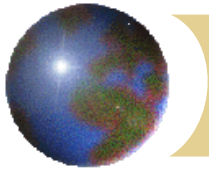


The set up

- ✚ We generated 3 million stock price paths for training using the process assumed by Black-Scholes-Merton
- ✚ We assumed that a trader wishes to hedge a short position in 10 call options
- ✚ The options last 10 days
- ✚ Probability of exploration starts at 1 and has a decay factor of 0.999999
- ✚ Each day the trader can change her position so that 0, 1, 2, ..., or 10 shares are held
- ✚ Hedging cost on day i is

$$H_i = N_i(S_{i+1} - S_i) - 10(c_{i+1} - c_i)$$

where N_i is the number of shares held, S_i is the share price at the beginning of Day i and c_i is the BSM call option price at the beginning of Day i



Test Set Results

- ⊕ The objective was to minimize the variance of total hedging costs
- ⊕ Our results were close to those given by delta hedging
- ⊕ The mean absolute difference between positions taken by the algorithm and those that would be taken with delta hedging when averaged across all hedging days was 0.33



Extensions

- ✚ Results can be extended to the situation where there are transactions costs (non-negligible bid-offer spreads) so that (a) delta hedging is not optimal and (b) the problem is a genuine multi-period one
- ✚ It is then necessary to trade off the mean cost of hedging with the variance of the cost of hedging
- ✚ A similar approach can be used to hedge volatility where transaction costs are high
- ✚ A mixture of processes can be used when the hedger is uncertain about the true process. (This is a device to ensure that the hedging strategy works reasonably well for all processes.)



Other Finance Applications

- ✚ Investing
- ✚ Private equity
- ✚ Hiring decisions
- ✚ Lending
- ✚ Identifying fraud
- ✚ Order execution
- ✚ Collateral management
- ✚ Fast derivative valuation