

Arbitrages, and pricing of stock options

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Arbitrages

Risk neutral measure

Black-Scholes formula for option pricing

- ▶ Bet on different events with each outcome paying a random return
- ▶ **Arbitrage**: It is possible to devise a betting strategy that **guarantees a positive return** no matter the combined outcome of the events
- ▶ Arbitrages often involve operating in two different markets

- ▶ Booker 1 \Rightarrow Phillies win pay 1.5:1, Phillies lose pay 3:1
- ▶ Bet x on Phillies and y against Phillies. Guaranteed Earnings?

$$\text{Phillies win: } 0.5x - y > 0 \Rightarrow x > 2y$$

$$\text{Phillies lose: } -x + 2y > 0 \Rightarrow x < 2y$$

- ▶ Arbitrage not possible. Notice that $1/(1.5) + 1/3 = 1$
- ▶ Booker 2 \Rightarrow Phillies win pay 1.4:1, Phillies lose pay 3.1:1
- ▶ Bet x on Phillies and y against Phillies. Guaranteed Earnings?

$$\text{Phillies win: } 0.4x - y > 0 \Rightarrow x > 2.5y$$

$$\text{Phillies lose: } -x + 2.1y > 0 \Rightarrow x < 2.1y$$

- ▶ Arbitrage not possible. Notice that $1/(1.4) + 1/(3.1) > 1$

- ▶ First condition on Booker 1 and second on Booker 2 are compatible
- ▶ Bet x on Phillies on Booker 1, y against Phillies on Booker 2
- ▶ Guaranteed earnings possible. Make $y = 1,000$, $x = 2,066$

$$\text{Phillies win: } 0.5(2,066) - 1,000 = 33$$

$$\text{Phillies loose: } -2066 + 2.1(1000) = 34$$

- ▶ Notice that $1/(1.5) + 1/(3.1) < 1$
- ▶ If you plan on doing this, do it on, e.g., currency exchange markets

- ▶ Let **events** on which bets are posted be $k = 1, 2, \dots, K$
- ▶ Let $j = 1, 2, \dots, J$ index possible **joint outcomes**
 - ▶ Joint realizations, also called “world realization”, or “world outcome”
- ▶ If **world outcome is j** , **event k** yields return r_{jk} per unit invested (bet)
- ▶ Do not define probability p_j of outcome j
- ▶ **Invest (bet) x_k in outcome k** \Rightarrow **return for world j** is $x_k r_{jk}$
- ▶ Bets x_k can be positive ($x_k > 0$) or negative ($x_k < 0$)
 - \Rightarrow Positive = regular bet. Negative = short bet

- ▶ Total return $\Rightarrow \sum_{k=1}^K x_k r_{jk} = \mathbf{x}^T \mathbf{r}_j$

- ▶ Vectors of **returns for outcome j** $\Rightarrow \mathbf{r}_j := [r_{j1}, \dots, r_{jK}]^T$ (given)
- ▶ Vector of **bets** $\Rightarrow \mathbf{x}_j := [x_{j1}, \dots, x_{jK}]^T$ (controlled by gambler)

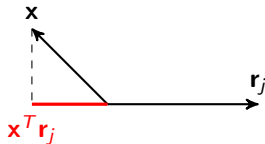
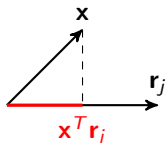
- ▶ Arbitrage is possible if there exists investment strategy \mathbf{x} such that

$$\mathbf{x}^T \mathbf{r}_j > 0, \quad \text{for all } j = 1, \dots, J$$

- ▶ Equivalently, arbitrage is possible if

$$\max_{\mathbf{x}} \left(\min_j (\mathbf{x}^T \mathbf{r}_j) \right) > 0$$

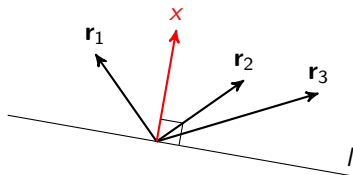
- ▶ Portfolio \mathbf{x} and returns \mathbf{r}_j are vectors in \mathbb{R}^K
- ▶ Earnings $\mathbf{x}^T \mathbf{r}_j$ are the inner product of \mathbf{x} and \mathbf{r}_j



- ▶ Earnings are positive if angle between \mathbf{x} and \mathbf{r}_j is less than $\pi/2$ (90°)

When is arbitrage possible?

- ▶ There is a line that leaves all \mathbf{r}_j vectors to one side

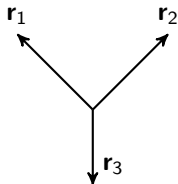


- ▶ Arbitrage possible
- ▶ Prob. vector $\mathbf{p} = [p_1, \dots, p_J]^T$ on world outcomes such that

$$\mathbb{E}_{\mathbf{p}}(\mathbf{r}) = \sum_{j=1}^J p_j \mathbf{r}_j = \mathbf{0}$$

- ▶ does **not** exist

- ▶ There is **not** a line that leaves all \mathbf{r}_j vectors to one side



- ▶ Arbitrage **not** possible
- ▶ There is prob. vector $\mathbf{p} = [p_1, \dots, p_J]^T$ on world outcomes such that

$$\mathbb{E}_{\mathbf{p}}(\mathbf{r}) = \sum_{j=1}^J p_j \mathbf{r}_j = \mathbf{0}$$

- ▶ Think of p_j as scaling factor

- ▶ Have “proved” following result, called arbitrage theorem

Theorem

Given vectors of returns \mathbf{r}_j , associated with random outcome $j = 1, \dots, J$ an *arbitrage is not possible* if and only if there exist a probability vector \mathbf{p} such that $\mathbb{E}_{\mathbf{p}}(\mathbf{r}) = \mathbf{0}$. Equivalently,

$$\max_{\mathbf{x}} \left(\min_j (\mathbf{x}^T \mathbf{r}_j) \right) \leq 0 \quad \Leftrightarrow \quad \sum_{j=1}^J p_j \mathbf{r}_j = \mathbf{0}$$

- ▶ Prob. vector \mathbf{p} is **NOT** the prob. distribution of events $j = 1, \dots, J$

- ▶ Consider a stock price $X(nh)$ that follows a geometric random walk

$$X((n+1)h) = X(nh)e^{\sigma\sqrt{h}Y_n}$$

- ▶ where Y_n is a binary random variable with probability distribution

$$P[Y_n = 1] = \frac{1}{2} \left(1 + \frac{\mu}{\sigma} \sqrt{h} \right), \quad P[Y_n = -1] = \frac{1}{2} \left(1 - \frac{\mu}{\sigma} \sqrt{h} \right)$$

- ▶ Recall that as $h \rightarrow 0$, $X(nh)$ becomes geometric Brownian motion
- ▶ Are there arbitrage opportunities in the price of the stock?
⇒ Too general, let us consider a narrower problem

- ▶ Consider the following investment strategy (stock flip):
 - Buy:** Buy \$1 in stock at time 0 for price $X(0)$ per unit of stock
 - Sell:** Sell stock at time h for price $X(h)$ for unit of stock
- ▶ Cost of transaction is \$1. Units of stock purchased are $1/X(0)$
- ▶ Cash after selling stock is $X(h)/X(0)$
- ▶ Return on investment is $X(h)/X(0) - 1$
- ▶ There are two possible outcomes for the price of the stock at time h
- ▶ As per model we may have $Y_0 = 1$ or $Y_0 = -1$ respectively yielding

$$X(h) = X(0)e^{\sigma\sqrt{h}}, \quad X(h) = X(0)e^{-\sigma\sqrt{h}}$$

- ▶ Possible returns are therefore

$$r_1 = \frac{X(0)e^{\sigma\sqrt{h}}}{X(0)} - 1 = e^{\sigma\sqrt{h}} - 1, \quad r_2 = \frac{X(0)e^{-\sigma\sqrt{h}}}{X(0)} - 1 = e^{-\sigma\sqrt{h}} - 1$$

- ▶ One dollar at time h is not the same as 1 dollar at time 0
- ▶ Interest rate of a risk-free investment is α continuously compounded
- ▶ In practice, α is the money market rate
- ▶ Prices have to be compared at their present value
- ▶ The present value of $X(h)$ at time 0 is $X(h)e^{-\alpha h}$
- ▶ Then, return on investment is $e^{-\alpha h}X(h)/X(0) - 1$
- ▶ Present value of possible returns (whether $Y_0 = 1$ or $Y_0 = -1$) are

$$r_1 = \frac{e^{-\alpha h}X(0)e^{\sigma\sqrt{h}}}{X(0)} - 1 = e^{-\alpha h}e^{\sigma\sqrt{h}} - 1,$$

$$r_2 = \frac{e^{-\alpha h}X(0)e^{-\sigma\sqrt{h}}}{X(0)} - 1 = e^{-\alpha h}e^{-\sigma\sqrt{h}} - 1$$

- ▶ Arbitrage not possible if and only if there exists $0 \leq q \leq 1$ such that

$$qr_1 + (1 - q)r_2 = 0$$

- ▶ Arbitrage theorem in 1 dimension (only one bet, buy stock)
- ▶ Substituting r_1 and r_2 for their respective values

$$q \left(e^{-\alpha h} e^{\sigma\sqrt{h}} - 1 \right) + (1 - q) \left(e^{-\alpha h} e^{-\sigma\sqrt{h}} - 1 \right) = 0$$

- ▶ Can be easily solved for q . Expanding product and reordering terms

$$qe^{-\alpha h} e^{\sigma\sqrt{h}} + (1 - q)e^{-\alpha h} e^{-\sigma\sqrt{h}} = 1$$

- ▶ Multiplying by $e^{\alpha h}$ and grouping terms with a q factor

$$q \left(e^{\sigma\sqrt{h}} - e^{-\sigma\sqrt{h}} \right) = e^{\alpha h} - e^{-\sigma\sqrt{h}}$$

- ▶ Solving for q finally yields $\Rightarrow q = \frac{e^{\alpha h} - e^{-\sigma\sqrt{h}}}{e^{\sigma\sqrt{h}} - e^{-\sigma\sqrt{h}}}$
- ▶ For small h we have $e^{\alpha h} \approx 1 + \alpha h$ and $e^{\pm\sigma\sqrt{h}} \approx 1 \pm \sigma\sqrt{h} + \sigma^2 h/2$
- ▶ Thus, the value of q as $h \rightarrow 0$ may be approximated as

$$q \approx \frac{1 + \alpha h - (1 - \sigma\sqrt{h} + \sigma^2 h/2)}{1 + \sigma\sqrt{h} - (1 - \sigma\sqrt{h})} = \frac{\sigma\sqrt{h} + (\alpha - \sigma^2/2) h}{2\sigma\sqrt{h}}$$
$$= \frac{1}{2} \left(1 + \frac{\alpha - \sigma^2/2}{\sigma} \sqrt{h} \right)$$

- ▶ Approximation proves that at least for small h $0 < q < 1$
 \Rightarrow Arbitrage not possible
- ▶ Also, suspiciously similar to probabilities of geometric random walk
 \Rightarrow Fundamental observation as we'll see next

Arbitrages

Risk neutral measure

Black-Scholes formula for option pricing

- ▶ Stock prices $X(t)$ follow geometric random walk (drift μ , variance σ^2)
- ▶ Risk free investment has return α (cost of money, money market)
- ▶ Arbitrage is not possible in **stock flips** if there is $0 \leq q \leq 1$ such that

$$q = \frac{e^{\alpha h} - e^{-\sigma\sqrt{h}}}{e^{\sigma\sqrt{h}} - e^{-\sigma\sqrt{h}}}$$

- ▶ Notice that q satisfies the equation (which we'll use later on)

$$qe^{\sigma\sqrt{h}} + (1 - q)e^{-\sigma\sqrt{h}} = e^{\alpha h}$$

- ▶ Can we have arbitrage using a **more complex set of possible bets**?

- ▶ Consider the following general investment strategy:
 - Observe:** Observe the stock price at times $h, 2h, \dots, nh$
 - Compare:** Is $X(h) = x_1, X(2h) = x_2, \dots, X(nh) = x_n$?
 - Buy:** If above answer is yes, buy stock at price $X(nh)$
 - Sell:** Sell stock at time mh for price $X(mh)$
- ▶ Possible bets are the observed values of the stock x_1, x_2, \dots, x_l
 - ⇒ There are 2^n possible bets
- ▶ Possible outcomes are value at time mh and observed values
 - ⇒ There are 2^m possible outcomes

- ▶ Bet 1 = n price increases, bet 2 = price increases in $1, \dots, n-1$ and price decrease in $n \dots$
- ▶ For each bet we have 2^{m-n} possible outcomes: $m-n$ price increases, price increases in $n+1, \dots, m-1$ and price decrease in $m \dots$

| | $X(h)$ | $X(2h)$ | $X(3h)$ | ... | $X(nh)$ | | $X((n+1)h)$ | $X((n+2)h)$ | ... | $X(mh)$ |
|-----------|-----------------------|------------------------|------------------------|-----|---------------------------|---|----------------------------|-----------------------------|-----|--------------------------------|
| bet 1 | $e^{\sigma\sqrt{h}}$ | $e^{2\sigma\sqrt{h}}$ | $e^{3\sigma\sqrt{h}}$ | ... | $e^{n\sigma\sqrt{h}}$ | ↘ | $X(nh)e^{\sigma\sqrt{h}}$ | $X(nh)e^{2\sigma\sqrt{h}}$ | ... | $X(nh)e^{m\sigma\sqrt{h}}$ |
| bet 2 | $e^{\sigma\sqrt{h}}$ | $e^{2\sigma\sqrt{h}}$ | $e^{3\sigma\sqrt{h}}$ | ... | $e^{(n-2)\sigma\sqrt{h}}$ | ↘ | $X(nh)e^{\sigma\sqrt{h}}$ | $X(nh)e^{2\sigma\sqrt{h}}$ | ... | $X(nh)e^{(m-2)\sigma\sqrt{h}}$ |
| bet 2^n | $e^{-\sigma\sqrt{h}}$ | $e^{-2\sigma\sqrt{h}}$ | $e^{-3\sigma\sqrt{h}}$ | ... | $e^{-n\sigma\sqrt{h}}$ | ↘ | $X(nh)e^{-\sigma\sqrt{h}}$ | $X(nh)e^{-2\sigma\sqrt{h}}$ | ... | $X(nh)e^{-m\sigma\sqrt{h}}$ |

outcomes per each bet

- ▶ Figure assumes $X(0) = 1$ for simplicity

- ▶ Define the prob. distribution \mathbf{q} over possible outcomes as follows
- ▶ Start with a sequence of independent identically distributed Y_n
- ▶ Each element Y_n is a binary random variable with probabilities

$$P[Y_n = 1] = q, \quad P[Y_n = -1] = 1 - q$$

- ▶ With $q = (e^{\alpha h} - e^{-\sigma\sqrt{h}})/(e^{\sigma\sqrt{h}} - e^{-\sigma\sqrt{h}})$ as in slide 16
- ▶ Joint prob. distribution \mathbf{q} on $X(h), X(2h), \dots, X((n+m)h)$ outcomes obtained through transformation

$$X((n+1)h) = X(nh)e^{\sigma\sqrt{h}Y_n}$$

- ▶ Notice once more that this is **NOT** the prob. distribution of $X(nh)$
- ▶ Will show that expected value of earnings with respect to \mathbf{q} is null
⇒ Thus, arbitrages are not possible

- ▶ Consider a time 0 unit investment in given arbitrary outcome
- ▶ Stock units purchased depend on the price $X(nh)$ at buying time

$$\text{Units bought} = \frac{1}{X(nh)e^{-\alpha nh}}$$

- ▶ Have corrected $X(nh)$ to express it in time 0 values
- ▶ Cash after selling stock given by price $X(mh)$ at sell time $m + n$
- ▶ Expressed in time 0 values

$$\text{Cash after sell} = \frac{X(mh)e^{-\alpha mh}}{X(nh)e^{-\alpha nh}}$$

- ▶ Return is then $\Rightarrow r(X(h), \dots, X(mh)) = \frac{X(mh)e^{-\alpha mh}}{X(nh)e^{-\alpha nh}} - 1$
- ▶ Depends on $X(mh)$ and $X(nh)$ only

- ▶ Consider expected value of all possible returns with respect to \mathbf{q}

$$\mathbb{E}_{\mathbf{q}} [r(X(h), \dots, X(mh))] = \mathbb{E}_{\mathbf{q}} \left[\frac{X(mh)e^{-\alpha mh}}{X(nh)e^{-\alpha nh}} - 1 \right]$$

- ▶ Condition on **observed values** $X(h), \dots, X(nh)$

$$\begin{aligned} \mathbb{E}_{\mathbf{q}} [r(X(h), \dots, X(mh))] \\ = \mathbb{E}_{\mathbf{q}(1:n)} \left[\mathbb{E}_{\mathbf{q}(n+1:m)} \left[\frac{X(mh)e^{-\alpha mh}}{X(nh)e^{-\alpha nh}} - 1 \mid X(h), \dots, X(nh) \right] \right] \end{aligned}$$

- ▶ In innermost expectation $X(nh)$ is given. Furthermore, process $X(t)$ is Markov, thus conditioning on $X(h), \dots, X((n-1)h)$ is irrelevant. Thus

$$\mathbb{E}_{\mathbf{q}} [r(X(h), \dots, X(mh))] = \mathbb{E}_{\mathbf{q}(1:n)} \left[\frac{\mathbb{E}_{\mathbf{q}(n+1:m)} [X(mh) \mid X(nh)] e^{-\alpha mh}}{X(nh)e^{-\alpha nh}} - 1 \right]$$

- ▶ Need to find **expectation of future value** $\mathbb{E}_{\mathbf{q}(n+1:m)} [X(mh) | X(nh)]$
- ▶ From recursive relation for $X(nh)$ in terms of Y_n sequence

$$\begin{aligned} X(mh) &= X((m-1)h) e^{\sigma\sqrt{h}Y_{m-1}} \\ &= X((m-2)h) e^{\sigma\sqrt{h}Y_{m-1}} e^{\sigma\sqrt{h}Y_{m-2}} \\ &\vdots \\ &= X(nh) e^{\sigma\sqrt{h}Y_{m-1}} e^{\sigma\sqrt{h}Y_{m-2}} \dots e^{\sigma\sqrt{h}Y_{n+1}} \end{aligned}$$

- ▶ All the Y_n are independent. Then, upon taking expected value

$$\mathbb{E}_{\mathbf{q}(n+1:m)} [X(mh) | X(nh)] = X(nh) \mathbb{E} \left[e^{\sigma\sqrt{h}Y_{m-1}} \right] \mathbb{E} \left[e^{\sigma\sqrt{h}Y_{m-2}} \right] \dots \mathbb{E} \left[e^{\sigma\sqrt{h}Y_{n+1}} \right]$$

- ▶ Need to determine **expectation of relative price increase** $\mathbb{E} \left[e^{\sigma\sqrt{h}Y_n} \right]$

- ▶ The expected value of the relative price increase $\mathbb{E} \left[e^{\sigma\sqrt{h}Y_n} \right]$ is

$$\mathbb{E} \left[e^{\sigma\sqrt{h}Y_n} \right] = e^{\sigma\sqrt{h}} \Pr[Y_n = 1] + e^{-\sigma\sqrt{h}} \Pr[Y_n = -1]$$

- ▶ According to definition of measure \mathbf{q} , it holds

$$\Pr[Y_n = 1] = q, \quad \Pr[Y_n = -1] = 1 - q$$

- ▶ Substituting in expression for $\mathbb{E} \left[e^{\sigma\sqrt{h}Y_n} \right]$

$$\mathbb{E} \left[e^{\sigma\sqrt{h}Y_n} \right] = e^{\sigma\sqrt{h}} q + e^{-\sigma\sqrt{h}} (1 - q) = e^{\alpha h}$$

- ▶ where last equality follows from definition of probability q
- ▶ Reweave the quilt \Rightarrow use expected relative price increase to compute expected future value to find expected return

- ▶ Substitute expected relative price increase into expression for expected future value

$$\mathbb{E}_{\mathbf{q}(n+1:m)} [X(mh) \mid X(nh)] = X(nh) e^{\alpha h} e^{\alpha h} \dots e^{\alpha h} = X(nh) e^{\alpha(m-n)h}$$

- ▶ Substitute result into expression for expected return

$$\mathbb{E}_{\mathbf{q}} [r(X(h), \dots, X(mh))] = \mathbb{E}_{\mathbf{q}(1:n)} \left[\frac{X(nh) e^{\alpha(m-n)h} e^{-\alpha mh}}{X(nh) e^{-\alpha nh}} - 1 \right]$$

- ▶ Exponentials cancel each other, finally yielding

$$\mathbb{E}_{\mathbf{q}} [r(X(h), \dots, X(mh))] = \mathbb{E}_{\mathbf{q}(1:n)} [1 - 1] = 0$$

- ▶ Arbitrage not possible in any trading strategy if $0 \leq q \leq 1$ exists

- ▶ Stock prices follow a geometric Brownian motion, i.e.,

$$X(t) = X(0)e^{Y(t)}$$

- ▶ with $Y(t)$ Brownian motion with drift μ and variance σ^2
- ▶ What is the no arbitrage condition?
- ▶ Approximate geometric Brownian motion by geometric random walk
- ▶ No arbitrage measure \mathbf{q} exists for geometric random walk
 - ▶ This requires h sufficiently small
 - ▶ Notice that **prob. distribution $\mathbf{q} = \mathbf{q}(h)$ is a function of h**
- ▶ Approximation arbitrarily accurate by letting $h \rightarrow 0$
- ▶ Existence of the **prob. distribution $\mathbf{q} := \lim_{h \rightarrow 0} \mathbf{q}(h)$** proves that arbitrages are not possible in stock trading

- ▶ Recall that as $h \rightarrow 0 \Rightarrow q \approx \frac{1}{2} \left(1 + \frac{\alpha - \sigma^2/2}{\sigma} \sqrt{h} \right)$
- ▶ And consequently $\Rightarrow (1 - q) \approx \frac{1}{2} \left(1 - \frac{\alpha - \sigma^2/2}{\sigma} \sqrt{h} \right)$
- ▶ Thus, **measure $\mathbf{q} := \lim_{h \rightarrow 0} \mathbf{q}(h)$ is geometric Brownian motion**
 - \Rightarrow **Variance** $\Rightarrow \sigma^2$ (same as stock price)
 - \Rightarrow **Drift** $\Rightarrow \alpha - \sigma^2/2$
- ▶ Measure showing arbitrage not possible is a geometric random walk
- ▶ Which is also the way stock prices evolve
- ▶ Furthermore, the variance is the same as that of stock prices
- ▶ The **drifts are different** $\Rightarrow \mu$ for stocks and $\alpha - \sigma^2/2$ for no arbitrage

- ▶ Compute expected return on an investment on stock $X(t)$
- ▶ Buy 1 share of stock at time 0. Cash invested $\Rightarrow X(0)$
- ▶ Sell stock at time t . Cash after sell $\Rightarrow X(t)$
- ▶ Expected value of cash after sell given $X(0)$ is

$$\mathbb{E} [X(t) \mid X(0)] = X(0)e^{(\mu + \sigma^2/2)t}$$

- ▶ Alternatively, invest $X(0)$ risk free in the money market
- ▶ Guaranteed cash at time t is $X(0)e^{\alpha t}$
- ▶ Invest in stock only if $\mu + \sigma^2/2 > \alpha \Rightarrow$ risk premium

- ▶ Compute expected return as if \mathbf{q} were the actual distribution
 - ▶ And recall that \mathbf{q} is **NOT** the actual distribution
- ▶ As before, cash invested is $X(0)$ and cash after sale is $X(t)$
- ▶ Expected cash value is different because prob. distribution is different

$$\mathbb{E}_{\mathbf{q}} [X(t) | X(0)] = X(0)e^{(\alpha - \sigma^2/2 + \sigma^2/2)t} = X(0)e^{\alpha t}$$

- ▶ **Same return as risk free investment** regardless of parameters' values
- ▶ Measure \mathbf{q} is called risk neutral measure
- ▶ Risky stock investments yield same return as risk free investments
- ▶ “Alternate universe” in which investors do not demand risk premiums
- ▶ **Pricing of derivatives, e.g., options, is always based on expected returns with respect to risk neutral valuation** (pricing in alternate universe)
- ▶ Basis for Black-Scholes. More later

Arbitrages

Risk neutral measure

Black-Scholes formula for option pricing

- ▶ An option is a contract to buy shares of a stock at a future time
- ▶ Strike time t = Convened time for stock purchase
- ▶ Strike price K = Price at which stock is purchased at strike time
- ▶ At time t , option holder may decide to
 - ⇒ Buy a stock at strike price K = exercise the option
 - ⇒ Do not exercise the option
- ▶ May buy option at time 0 for price c
- ▶ How do we **determine the option's worth**, i.e., price c , at time 0 ?
- ▶ Answer given by Black-Scholes formula for option pricing

- ▶ Let $e^{\alpha t}$ be the compounding of a risk free investment
- ▶ Let $X(t)$ be the stock's price at time t
- ▶ Price modeled as geometric Brownian motion, drift μ , variance σ^2
- ▶ Risk neutral measure \mathbf{q} is also a geometric Brownian motion
⇒ Variance σ^2 and drift $\alpha - \sigma^2/2$

- ▶ At time t , the option's worth depends on the stock's price $X(t)$
- ▶ If stock's price smaller or equal than strike price $\Rightarrow X(t) \leq K$
 \Rightarrow Option is worthless (better to buy stock at current price)
- ▶ Since had paid c for the option at time 0, lost c on this investment
 \Rightarrow return on investment is $r = -c$
- ▶ If stock's price larger than strike price $\Rightarrow X(t) > K$
 \Rightarrow Exercise option and realize a gain of $X(t) - K$
- ▶ To obtain return express as time 0 values and subtract c

$$r = e^{-\alpha t}(X(t) - K) - c$$

- ▶ May combine both in single equation $\Rightarrow r = e^{-\alpha t}(X(t) - K)^+ - c$
- ▶ $(\cdot)^+$ denotes projection on positive reals

- ▶ Consider mixed positions on stocks and options
- ▶ Is there a position guaranteeing positive return, i.e., an arbitrage?
- ▶ Assume expected return under risk neutral measure is nonzero

$$\mathbb{E}_{\mathbf{q}}[r] = \mathbb{E}_{\mathbf{q}} \left[e^{-\alpha t} (X(t) - K)^+ - c \right] \neq 0$$

- ▶ Then, an arbitrage is possible according to arbitrage theorem
- ▶ If expected return under risk neutral measure is zero

$$\mathbb{E}_{\mathbf{q}}[r] = \mathbb{E}_{\mathbf{q}} \left[e^{-\alpha t} (X(t) - K)^+ - c \right] = 0$$

- ▶ Then, no arbitrage is possible according to arbitrage theorem
- ▶ **Select options price c to prevent arbitrage opportunities**

- ▶ To have no arbitrage, must select option's price c so that

$$\mathbb{E}_{\mathbf{q}} \left[e^{-\alpha t} (X(t) - K)^+ - c \right] = 0$$

- ▶ where expectation is with respect to risk neutral measure
- ▶ From above condition, the no-arbitrage price of the option is

$$c = e^{-\alpha t} \mathbb{E}_{\mathbf{q}} \left[(X(t) - K)^+ \right]$$

- ▶ Source of Black-Scholes formula for option valuation
- ▶ Rest of derivation is just evaluation of expected value
- ▶ Same argument used to price any derivative of the stock's price

- ▶ Let us evaluate expectation to compute option's price c
- ▶ Prices follow a geometric random walk $\Rightarrow X(t) = X_0 e^{Y(t)}$
- ▶ X_0 = price at time 0,
- ▶ $Y(t)$ random walk with drift parameter μ and variance parameter σ^2
- ▶ Can rewrite no arbitrage condition as

$$c = e^{-\alpha t} \mathbb{E}_{\mathbf{q}} \left[\left(X_0 e^{Y(t)} - K \right)^+ \right]$$

- ▶ $Y(t)$ random walk. Then, in particular, $Y(t) \sim \mathcal{N}(\mu t, t\sigma^2)$

$$c = e^{-\alpha t} \frac{1}{\sqrt{2\pi t\sigma^2}} \int_{-\infty}^{\infty} (X_0 e^y - K)^+ e^{-(y-\mu t)^2/(2t\sigma^2)} dy$$

- ▶ Note that $(X_0 e^{Y(t)} - K)^+ = 0$ for all values $Y(t) \leq \log(K/X_0)$
- ▶ Because integrand is null for $Y(t) \leq \log(K/X_0)$ can write

$$c = e^{-\alpha t} \frac{1}{\sqrt{2\pi t\sigma^2}} \int_{\log(K/X_0)}^{\infty} (X_0 e^y - K) e^{-(y-\mu t)^2/(2t\sigma^2)} dy$$

- ▶ Change of variables $z = (y - \mu t)/\sqrt{t\sigma^2}$. Associated replacements

Variable: $y \Rightarrow \sqrt{t\sigma^2}z + \mu t$

Differential: $dy \Rightarrow \sqrt{t\sigma^2} dz$

Integration limit: $\log(K/X_0) \Rightarrow a := \frac{\log(K/X_0) - \mu t}{\sqrt{t\sigma^2}}$

- ▶ Option price then given by

$$c = e^{-\alpha t} \frac{1}{\sqrt{2\pi}} \int_a^{\infty} (X_0 e^{\sqrt{t\sigma^2}z + \mu t} - K) e^{-z^2/2} dz$$

- ▶ Separate in two integrals $c = e^{-\alpha t}(I_1 - I_2)$ where

$$I_1 := \frac{1}{\sqrt{2\pi}} \int_a^\infty X_0 e^{\sqrt{t}\sigma^2 z + \mu t} e^{-z^2/2} dz$$

$$I_2 := \frac{K}{\sqrt{2\pi}} \int_a^\infty e^{-z^2/2} dz$$

- ▶ Gaussian Q function (ccdf of normal RV with mean 0 and variance 1)

$$Q(x) := \frac{1}{\sqrt{2\pi}} \int_x^\infty e^{-z^2/2} dz$$

- ▶ Comparing last two equations we have $I_2 = KQ(a)$
- ▶ I_1 requires some more work

- ▶ Reorder terms in integral I_2

$$I_1 := \frac{1}{\sqrt{2\pi}} \int_a^\infty X_0 e^{\sqrt{t\sigma^2}z + \mu t} e^{-z^2/2} dz = \frac{X_0 e^{\mu t}}{\sqrt{2\pi}} \int_a^\infty e^{\sqrt{t\sigma^2}z - z^2/2} dz$$

- ▶ The exponent can be written as a square minus a “constant” (no z)

$$-\left(z - \sqrt{t\sigma^2}\right)^2 / 2 + t\sigma^2 / 2 = -z^2 / 2 + \sqrt{t\sigma^2}z - t\sigma^2 / 2 + t\sigma^2 / 2$$

- ▶ Substituting the latter into I_1 yields

$$I_1 = \frac{X_0 e^{\mu t}}{\sqrt{2\pi}} \int_a^\infty e^{-(z - \sqrt{t\sigma^2})^2 / 2 + t\sigma^2 / 2} dz = \frac{X_0 e^{\mu t + t\sigma^2 / 2}}{\sqrt{2\pi}} \int_a^\infty e^{-(z - \sqrt{t\sigma^2})^2 / 2} dz$$

- ▶ Change of variables $u = z - \sqrt{t\sigma^2} \Rightarrow du = dz$ and integration limit

$$a \Rightarrow b := a - \sqrt{t\sigma^2} = \frac{\log(K/X_0) - \mu t}{\sqrt{t\sigma^2}} - \sqrt{t\sigma^2}$$

- ▶ Implementing change of variables in I_1

$$I_1 = \frac{X_0 e^{\mu t + t\sigma^2/2}}{\sqrt{2\pi}} \int_b^\infty e^{u^2/2} du = X_0 e^{\mu t + t\sigma^2/2} Q(b)$$

- ▶ Putting together results for I_1 and I_2

$$c = e^{-\alpha t}(I_1 - I_2) = e^{-\alpha t} X_0 e^{\mu t + t\sigma^2/2} Q(b) - e^{-\alpha t} K Q(a)$$

- ▶ For non-arbitrage stock prices $\Rightarrow \alpha = \mu + \sigma^2/2$
- ▶ Substitute to obtain Black-Scholes formula

- ▶ Black-Scholes formula for option pricing

$$c = X_0 Q(b) - e^{-\alpha t} K Q(a)$$

- ▶ Where $\Rightarrow a := \frac{\log(K/X_0) - \mu t}{\sqrt{t\sigma^2}}$ and $b := a - \sqrt{t\sigma^2}$
- ▶ Note further that $\mu = \alpha - \sigma^2/2$. Can then write a as

$$a = \frac{\log(K/X_0) - (\alpha - \sigma^2/2) t}{\sqrt{t\sigma^2}}$$

- ▶ X_0 = stock price at time 0, c = option cost at time 0,
- ▶ K = option's strike price, t = option's strike time
- ▶ α = benchmark risk-free rate of return (cost of money)
- ▶ σ^2 = volatility of stock
- ▶ Black-Scholes formula independent of stock's mean tendency μ