

Outline

- We will now approach the option pricing problem from a probabilistic angle. You can think of this as an extension of our discrete option pricing problem.
- Some problems are a lot simpler when we consider the probability approach and it will finally link us with Monte Carlo methods.
- In particular we will introduce the following ideas:
 - Martingales (discrete and continuous)
 - Changing of probability measure (Girsanov's theorem)
 - Uniqueness of change of measure
 - Uniqueness of risk-neutral or martingale measure and no arbitrage/complete markets
 - Option pricing under this approach

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0 0 0

- Stock price, S .
- Option price, V .
- Stock price, S , under risk-neutral measure Q when $R = 0$
- Bond price, B , under risk-neutral measure Q .
- The total received, X , in a gambling game where you toss a coin a finite number of times, receiving \$1 for each head and paying \$1 for each tail?

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Option is a martingale

- For any option payoff X the process $\eta_t = E_t^P(X)$ is always a P martingale, i.e. a martingale under any probability measure.
- To see that this is true use the fact that

$$E_s^P[E_t^P[X]] = E_s^P[X]$$

so if we condition on history up until s and then conditioning on the history until t is the same as conditioning originally up to time s . This is called the **tower law**.

- This is more obvious than it initially appears!

Quiz

What result would ensure that η_t was a martingale ($s > t$)?

- $\eta_t = \eta_s$.
- $\eta_t = E_t[\eta_s]$
- $X_t = E_t[X_s]$

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Replication strategy

- Now choose a probability measure Q such that Z (recall $Z_t = S_t/B_t$) is a Q martingale (this is just our standard risk neutral measure) then there is a Q martingale process $\eta_t = E_t^Q[X/B_T]$.
- There is then a previsible process (portfolio) ϕ such that

$$\eta_t = \eta_0 + \sum_{k=1}^t \phi_k \Delta Z_k$$

- So at time t you purchase ϕ_{t+1} units of stock and $\psi_{t+1} = \eta_t - \phi_{t+1}S_t/B_t$ of the cash bond. As we have done in our binomial model
- At time 0, $\Pi_0 = \phi_1 S_0 + \psi_1 B_0 = \eta_0 = E_0^Q[X/B_T]$.
- Then the portfolio rolls forward, consider time = 1. Now we have

$$\begin{aligned} \Pi_1 &= \phi_1 S_1 + \psi_1 B_1 \\ &= B_1(\eta_0 + \phi_1(S_1/B_1 - S_0/B_0)) \\ &= B_1(\eta_0 + \phi_1 \Delta Z_1) \\ &= B_1 \eta_1 \end{aligned}$$

Replication strategy

- This is the cost of the new portfolio as $B_1\eta_1 = \phi_2S_1 + \psi_2B_1$, which in turn will go on to be worth $B_2\eta_2$ at the end of the next period and so on.
- Then the portfolio rolls forward again, regardless of how S changes, until at expiry the portfolio is worth $B_T X / B_T = X$ (just as with the binomial model).
- So the current price of the option is the initial value of the replicating portfolio, $\Pi_0 = \eta_0 = E_0^Q[X/B_T]$ or the expected value of the option under the Q (risk-neutral) measure.
- And, in general, $\Pi_t = B_t\eta_t = B_tE_t^Q[X/B_T]$

Quiz

What does this look like?

- An arbitrage strategy
- A self financing portfolio
- An ito process

Self-financing

- What is the definition of self financing here?
- The portfolio is self financing if the closing (i.e time t) value of portfolio Π_{t-1} constructed at t is worth $\Pi_t = \phi_{t-1}S_t + \psi_{t-1}B_t$.
- The financing gap is given by

$$D_t = \Pi_t - \phi_{t-1}S_t - \psi_{t-1}B_t$$

or alternatively:

$$\Delta \Pi_{t-1} = \phi_{t-1} \Delta S_{t-1} + \psi_{t-1} \Delta B_{t-1} + D_t$$

and so the gap D_t is only zero if the changes in the strategy only come from changes in the stock and the bond prices (as we had before).

Quiz

What is true about the process V_t/B_t here?

- It has a positive drift
- It is a martingale
- It mean reverts

Aside: why do we have to make Z the martingale?

- Above, we made $Z = S/B$ the martingale and not S , itself. Why not choose S .
- It is also possible to determine a probability where the stock price is a martingale. Why do we not choose this probability for option valuation?
- Let's see what would happen here: the Binomial representation theorem gives us:

$$\eta_t = \eta_0 + \sum_{k=1}^t \phi_k \Delta S_k$$

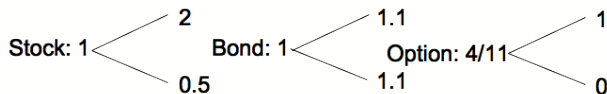
as S is now the martingale.

- Now, at $t = 0$, hold ϕ_1 of stock and $\psi_1 = \eta_0 - \phi_1 S_0$ in the risk-free asset, where $B_0 = 1$

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Easy numerical example

- Consider a stock, bond and option in a one period world



- Under the risk-neutral measures, $q = 0.4$ and $1 - q = 0.6$, we have:

$$E_0^Q \left[\frac{S(t)}{B(t)} \right] = \frac{0.4 \times 2 + 0.6 \times 0.5}{1.1} = \frac{1.1}{1.1} = 1 = \frac{S(0)}{B(0)}$$

thus S/B is a Martingale under the risk-neutral measure Q , where the unique nature of Q means that this also implies no arbitrage opportunities.

- We also have:

$$E_0^Q \left[\frac{V(t)}{B(t)} \right] = \frac{0.4 \times 1 + 0.6 \times 0}{1.1} = \frac{0.4}{1.1} = 4/11 = \frac{V(0)}{B(0)}$$

Easy numerical example

- So under our risk neutral measure V/B is also a martingale, and as we discussed in the fundamental theorem of finance, the risk-neutral measure exists and is unique if and only if we have complete markets and no arbitrage. Also recall that we can then apply the same measure to price all replicable assets (such as derivatives).
- What we intend to do is to extend this to continuous time in which case we can use martingale theory to derive the Black-Scholes PDE and calculate the value of European call and put options.

Quiz

What choice of q and $1 - q$ would make the **stock price** a martingale?

- It's not possible
- $q = 2/3, 1 - q = 1/3$
- $q = 0.5, 1 - q = 0.5$
- $q = 1/3, 1 - q = 2/3$

Continuous Martingales

- ... thus

$$E_t[W(T)] = W(t) + 0 = W(t)$$

- Finally. for any payoff X depending only upon events up to time T the process

$$\eta_t = E_t^P[X]$$

is a martingale.

- This follows from the tower law,

$$E_s^P [E_t^P[X]] = E_s^P[X]$$

as in the discrete setting.

- Stock price, S
- Option price, V
- Bond price, B
- $\sigma S dW$
- Stock price, S under risk-neutral measure Q
- Stock price, S under risk-neutral measure Q when $R = 0$
- Bond price, B under risk-neutral measure Q .

Driftless processes

- We need a way to determine if processes are martingales from our continuous SDEs.
- In our SDEs we had a drift term. Do martingales always have zero drift, and can martingales always be represented as an SDE with zero drift?
- If X_t is a martingale then it can be written as:

$$X_t = X_0 + \int_0^t \phi_s dW_s$$

where W is a Brownian motion and so $dX_t = \phi_t dW_t$.

- In general we find that if X is a stochastic process with volatility σ_t , $dX_t = \mu_t dt + \sigma_t dW_t$ then

X is a martingale $\Leftrightarrow X$ is driftless, $\mu_t = 0$

Continuous portfolios

- Now, with analogy to the binomial model, we will define a portfolio. A portfolio is a pair of processes ϕ_t and ψ_t which describe the number of units of stock and bond which we are holding at time t . The stock component, ϕ should be pre-visible, depending upon information available at time t .
- The self financing condition will be the same as in the PDE derivation:

$$\Delta \Pi_t = \phi_t \Delta S_t + \psi_t \Delta B_t$$

- A replicating strategy for a payoff X is a self financing portfolio (ϕ, ψ) such that $\Pi_T = \phi_T S_T + \psi_T B_T = X$.
- To progress further we need a new theorem.

Example of part 1

- Suppose we have a process:

$$dX_t = \mu_t dt + \sigma_t dW_t$$

where W is a Brownian motion under P and we wished to change the measure so that the drift was ν_t instead.

- Under Q we have

$$dX_t = \mu_t dt - \gamma \sigma_t dt + \sigma_t d\tilde{W}_t$$

and so if we choose:

$$\gamma = \frac{\mu_t - \nu_t}{\sigma_t}$$

then

$$dX_t = \nu_t dt + \sigma d\tilde{W}_t$$

Quiz

If we wished dX_t to have zero drift in the above example, what should γ equal?

- $\gamma = 0$.
- $\gamma = \mu_t$.
- $\gamma = \mu_t / \sigma_t$
- $\gamma = (\mu_t - r_t) / \sigma_t$

Quiz

What is S_T under Q , here?

- $S_T = S_0$
- $S_T = S_0 \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)T + \sigma W(T)\right)$
- $S_T = S_0 \exp\left(\left(r - \frac{1}{2}\sigma^2\right)T + \sigma W(T)\right)$
- $S_T = S_0 \exp\left(-\frac{1}{2}\sigma^2 T + \sigma W(T)\right)$

Full derivation ($r \neq 0$)

- Using Girsanov's theorem, choosing $\gamma_t = (\mu - r)/\sigma$ makes Z_t into a martingale:

$$dZ_t = \sigma Z_t d\tilde{W}_t$$

- Now consider the discounted claim process $\eta_t = E_t^Q \left[\frac{X}{B_T} \right]$ which is also a Q martingale.
- Now we need to try to form a replicating portfolio. Our martingale representation theorem tells us that $d\eta_t = \phi_t dZ_t$ so let's hold ϕ_t of the stock at time t and $\psi_t = \eta_t - \phi_t Z_t$ of the bond.
- This gives that

$$\Pi_t = \phi_t S_t + \psi_t B_t = B_t \eta_t$$

Quiz

What process does the stock price follow under the new measure, Q ?

- $dS_t = \sigma S_t d\tilde{W}_t$
- $dS_t = (r - \frac{1}{2}\sigma^2)S_t dt + \sigma S_t d\tilde{W}_t$
- $dS_t = rS_t dt + \sigma S_t d\tilde{W}_t$
- $dS_t = (\mu - \frac{1}{2}\sigma^2)S_t dt + \sigma S_t d\tilde{W}_t$

Quiz

What is does γ_t resemble?

- Stock beta.
- Sharpe ratio.
- Portfolio weight.
- Risk premium.

Full derivation ($r \neq 0$)

- But we know that at all points $\Pi_t = \eta_t B_t$ then:

$$\begin{aligned} d\Pi_t &= B_t d\eta_t + \eta_t dB_t \\ &= \phi_t B_t dZ_t + \eta_t dB_t \\ &= \phi_t B_t dZ_t + (\phi_t Z_t + \psi_t) dB_t \\ &= \phi_t (B_t dZ_t + Z_t dB_t) + \psi_t dB_t \end{aligned}$$

- But $(B_t dZ_t + Z_t dB_t) = d(B_t Z_t) = dS_t$ and so,

$$d\Pi_t = \phi_t dS_t + \psi_t dB_t$$

and so the portfolio is self-financing.

Valuing a call option

- As we know the pdf of the Normal distribution we can write

$$V_0 = \frac{e^{-rT}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-x^2/2} \max \left[\left\{ S_0 \exp \left(rT - \frac{1}{2} \sigma^2 T + \sigma \sqrt{T} x \right) - K, 0 \right\} \right] dx$$

- The integrand is zero unless

$$x \geq \frac{\ln(K/S_0) + \frac{1}{2} \sigma^2 T - rT}{\sigma \sqrt{T}} (= L)$$

thus,

$$\begin{aligned} V_0 &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_L^{\infty} e^{-x^2/2} S_0 \exp \left(rT - \frac{1}{2} \sigma^2 T + \sigma \sqrt{T} x \right) dx \\ &\quad - \frac{e^{-rT}}{\sqrt{2\pi}} \int_L^{\infty} e^{-x^2/2} K dx \\ &= A(S_0, t) - B(S_0, t) \end{aligned}$$

Call option valuation: $B(S_0, t)$

- Taking $B(S_0, t)$ first

$$B(S_0, t) = e^{-rT} K \frac{1}{\sqrt{2\pi}} \int_L^\infty e^{-\frac{x^2}{2}} dx$$

- This is just a constant term multiplied by the upper tail of a Normal distribution. Thus,

$$\begin{aligned} -B(S_0, t) &= -e^{-rT} K (1 - N(L)) \\ &= -e^{-rT} K N(-L) \\ &= -e^{-rT} K N(d_2) \end{aligned}$$

where

$$d_2 = -L = \frac{\ln(S_0/K) + (r - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}$$

Call option valuation: $A(S_0, t)$

- Next,

$$\begin{aligned} A(S_0, t) &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_L^\infty S_0 e^{(r-\frac{1}{2}\sigma^2)T + \sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx \\ &= \frac{S_0}{\sqrt{2\pi}} \int_L^\infty e^{-\frac{1}{2}\sigma^2 T + \sigma\sqrt{T}x - \frac{x^2}{2}} dx \\ &= \frac{S_0}{\sqrt{2\pi}} \int_L^\infty e^{-(x-\sigma\sqrt{T})^2/2} dx \end{aligned}$$

- Introduce a new variable $y = x - \sigma\sqrt{T}$ then we have

$$A(S_0, t) = \frac{S_0}{\sqrt{2\pi}} \int_{I - \sigma\sqrt{T}}^{\infty} e^{-y^2/2} dy.$$

Call option valuation: $A(S_0, t)$

- This is just a constant term multiplied by the upper tail of a Normal distribution. Thus,

$$\begin{aligned} A(S_0, t) &= S_0(1 - N(L - \sigma\sqrt{T})) \\ &= S_0N(-L + \sigma\sqrt{T}) \\ &= S_0N(d_1) \end{aligned}$$

where

$$d_1 = -L + \sigma\sqrt{T} = \frac{\ln(S_0/K) + (r + \frac{1}{2}\sigma^2)(T - t)}{\sigma\sqrt{(T - t)}}$$

Call option valuation

- This then gives us the Black-Scholes-Merton formula at a general time t

$$\begin{aligned} V_t &= A(S, t) - B(S, t) \\ &= S_t N(d_1) - e^{-r(T-t)} K N(d_2) \end{aligned}$$

where

$$\begin{aligned} d_1 &= \frac{\ln(S_t/K) + (r + \frac{1}{2}\sigma^2)(T - t)}{\sigma\sqrt{(T - t)}} \\ d_2 &= \frac{\ln(S_t/K) + (r - \frac{1}{2}\sigma^2)(T - t)}{\sigma\sqrt{(T - t)}} = d_1 - \sigma\sqrt{T - t} \end{aligned}$$

- I personally think that this derivation is far simpler than that from the PDE solution.

Very useful: Different numerares

- In the examples we have given the risk-free bond plays the role of the Numeraire asset. It is useful because it has no volatility but could we use any asset as the numeraire asset?
- In fact we can, we just have a revised pricing formula for any numeraire asset N_t so that now:

$$\frac{V_t}{N_t} = E_t^Q \left[\frac{V_T}{N_T} \right]$$

- This is very useful for foreign exchange problems and crucial for term structure work, as we will see later in the course.
- We will work through some examples of different numeraires in Problem Set 6.

Recap

- There are no arbitrage opportunities if and only if assets have a unique pricing measure.
- In a complete market, there are no arbitrage opportunities if and only if the ratio of any tradeable asset to the risk-free (or numeraire) asset is a martingale.
- Girsanov's theorem shows that there is a unique change of measure which made this ratio a martingale.
- All other replicable assets can be priced using the same unique measure as by the ratio of their price to the risk-free bond is also a martingale under the same measure.

Overview

- We have looked at derivative pricing from an entirely probability focussed approach. We have been able to derive the same results as before but it could be argued that the formulas are simpler.
- The crucial insight is that if there is no arbitrage then the current discounted value of an asset is its discounted expectation under the appropriate martingale measure.
- This gives us a simple formula for valuing options and a link back to our fundamental theorems. Next, we shall see some applications of this.

Advanced: Fundamental theorems of finance

- We have mixed together our fundamental theorem of finance and the martingale representation, we should be a little more careful.
- Our fundamental theorems of finance say that
 - Arbitrage \Leftrightarrow Risk-neutral measure.
 - Completeness \Leftrightarrow Unique measure.
- Our measure here is often called the equivalent martingale measure in continuous time finance. These are proved in the famous paper of Harrison and Pliska (1981). We omit the proofs but give a feel for the arguments next.

Advanced: Martingales and arbitrage

- Martingales are essentially the absence of arbitrage.
- To show this, assume that within our self financing replication strategy there is the possibility of an arbitrage.
- The portfolio value at time t is given by:

$$\Pi_t = \phi_t S_t + \psi_t B_t$$

where

$$d\Pi_t = \phi_t dS_t + \psi_t dB_t$$

and the discounted value of the portfolio is $\eta_t = \Pi_t / B_t$ and then

$$d\eta_t = \phi_t dZ_t$$

where $Z_t = S_t / B_t$.

Advanced: Martingales and arbitrage

- Suppose that the strategy starts with zero value, $\Pi_0 = 0$, and finished with a non-negative payoff, $\Pi_T \geq 0$. Is there really an arbitrage opportunity?
- As Z_t is a martingale then η_t is a martingale too and so:

$$E^Q[\eta_T] = E_0^Q[\eta_T] = \eta_0 = \Pi_0 = 0$$

and so η_T must take value 0, and thus $\Pi_T = 0$ and so if the Martingale measure Q exists then there are no arbitrage opportunities,

- We omit the other direction!

Advanced: Real probs \rightarrow R-N probs

- Consider a very simple two period random walk with, which does not necessarily recombine. To get from time 0 to time 2 there are four possible paths in the table below

Path	Probabilities
$\{0, 1, 2\}$	$p_1 p_2 = \pi_1$
$\{0, 1, 0\}$	$p_1 (1 - p_2) = \pi_2$
$\{0, -1, 0\}$	$(1 - p_1) p_3 = \pi_3$
$\{0, -1, -2\}$	$(1 - p_1) (1 - p_3) = \pi_4$

- The π values could denote the probability measure P .
- We could also denote the same paths using a different probability measure Q with path probabilities $\pi'_1, \pi'_2, \pi'_3, \pi'_4$

Advanced: Radon-Nikodym derivative

- The Radon-Nikodym derivative has some useful properties which are easiest to see in the discrete setting. Consider a claim (option) with payoffs x_i on path i then under P its expected value is:

$$E^P[X] = \sum_i \pi_i x_i$$

where i ranges over the four paths. The expectation with respect to Q is

$$E^Q[X] = \sum_i \pi'_i x_i = \sum_i \pi_i \left(\frac{\pi'_i}{\pi_i} x_i \right) = E^P \left[\frac{dQ}{dP} X \right]$$

- This simple result can be extended to give a useful general result:

$$E_0^Q[X_T] = E_0^P \left[\frac{dQ}{dP} X_T \right]$$

- At intermediate times $0 < t < T$ we need to know more about the Radon-Nikodym derivative, which we can do by converting it into a stochastic process, by letting the time horizon vary.

Advanced: Radon-Nikodym derivative

- Let ζ_t be the Radon-Nikodym derivative taken up to the horizon t . Thus in our simple example ζ_1 can take values π'_1/π_1 and $(1 - \pi'_1)/(1 - \pi_1)$.
- In general we can write:

$$\zeta_t = E_t^P \left[\frac{dQ}{dP} \right]$$

- ζ_t is very useful as it represents the amount of change of measure so far up to time t . So if we wanted to know $E^Q[X_t]$ it would be $E^P[\zeta_t X_t]$ and, more importantly for us, we have:

$$E_s^Q[X_t] = \frac{E_s^P[\zeta_t X_t]}{\zeta_s}$$

Advanced: Changing measure in continuous time

- We extend our nice discrete results to continuous time.
- We need to think a little more deeply about probability measures where we have continuous times and states.
- To do this we need all of the marginal distributions at each time t conditional on every history \mathcal{F}_s for all times $s < t$. We simplify it slightly by considering a large (but finite) set of times $0 = t_0, t_1, \dots, t_n - 1, t_n = T$ and points x_1, \dots, x_n .
- With one point, x and one time, t_1 we could use the normal density function:

$$f_1(x) = \frac{1}{\sqrt{2\pi t_1}} \exp\left(-\frac{x^2}{2t_1}\right)$$

- Extending it to different points at different times we get

$$f_n(x_1, \dots, x_n) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\Delta t_i}} \exp\left(-\frac{\Delta x_i^2}{2\Delta t_i}\right)$$

where $\Delta x_j = x_j - x_{j-1}$ and $\Delta t_j = t_j - t_{j-1}$.

Advanced: Changing measure in continuous time

- So in a non-elegant way we can define our Radon-Nikodym derivative as a limit:

$$\frac{dQ}{dP} = \lim_{n \rightarrow \infty} \frac{f_n^P(x_1, \dots, x_n)}{f_n^Q(x_1, \dots, x_n)}$$

- This still satisfies our earlier conditions:

$$E_0^Q[X_T] = E_0^P \left[\frac{dQ}{dP} X_T \right]$$

$$E_s^Q[X_t] = \frac{E_s^P[\zeta_t X_t]}{\zeta_s}$$

Advanced: Changing measure in continuous time: In practice

- Consider a Brownian motion, W_t , under the probability measure P . Now, as if by magic, define an equivalent probability measure, Q such that:

$$\frac{dQ}{dP} = \exp(-\gamma W_T - \frac{1}{2}\gamma^2 T)$$

- Before progressing, let's consider what a normally distributed random variable looks like. A random variable X is normally distributed (with mean μ and variance σ , if and only if

$$E[\exp(\theta X)] = \exp(\theta\mu + \frac{1}{2}\theta^2\sigma^2)$$

- Given that, let's try to calculate the expectation of θW_T under the measure Q :

$$\begin{aligned} E^Q[\exp(\theta W_T)] &= E^P\left[\exp\left(-\gamma W_T - \frac{1}{2}\gamma^2 T + \theta W_T\right)\right] \\ &= \exp\left(-\frac{1}{2}\gamma^2 T + \frac{1}{2}(\theta - \gamma)^2 T\right) \end{aligned}$$

Advanced: Changing measure in continuous time: In practice

or,

$$E^Q[\theta W_T] = \exp\left(-\theta\gamma T - \frac{1}{2}\theta^2 T\right)$$

Advanced: Girsanov's theorem (full version)

1. If W is a Brownian motion under the probability measure P , then there exists an equivalent measure Q such that $\tilde{W}_t = W_t + \gamma t$ is a Brownian motion.
2. Let W be a Brownian motion under a probability measure P , then under an equivalent measure:

$$Q(A) = E \left[1_A \frac{dQ}{dP} \right]$$

\tilde{W} is a Brownian motion with drift γ . Or $W = \tilde{W} + \gamma T$ is a Brownian motion. (A is an arbitrary set of paths (possible values of W_T) and 1_A is an indicator which takes the value 1 only when path A occurs.)

- In this case,

$$\frac{dQ}{dP} = \exp \left(- \int_0^T \gamma_t dW_t - \frac{1}{2} \int_0^T \gamma_t^2 dt \right)$$

Advanced: Example of part 2

- Consider switching between measures:

$$P(W_T < x) = P(N(0, 1) < x/\sqrt{T}) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x/\sqrt{T}} e^{-s^2/2} ds$$

transforming,

$$P(W_T < x) = \frac{1}{\sqrt{2\pi T}} \int_{-\infty}^x e^{-s^2/2T} ds$$

and so,

$$\begin{aligned} Q(W_T < x) = E[1_{W_T < x} M(T)] &= \frac{1}{\sqrt{2\pi T}} \int_{-\infty}^x e^{-\frac{s^2}{2T}} e^{-\frac{1}{2}\gamma^2 T - \gamma s} ds \\ &= \frac{1}{\sqrt{2\pi T}} \int_{-\infty}^x e^{-\frac{(s+\gamma T)^2}{2T}} ds \end{aligned}$$

- Simply changing $r = s + \gamma T$ shows that this is the same as

$$P(W_T < x + \gamma T).$$

Advanced: Example of part 2

- Thus the probability that $W_T < x$ in the new measure is equal to the probability that $W_T - \gamma T$ is less than x in the old measure. That is W_T is a Brownian motion with drift $-\gamma$ in the new measure, as required.