

# Valuing equity-linked death benefits and exotic contingent options in jump diffusion models

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## Equity-linked death benefit

- ▶  $x$  age at issue of policy (time 0)
- ▶  $T_x$  time of death
- ▶ payment at time  $T_x$
- ▶ depends on  $S(T_x)$ ,
- ▶ or more generally on  $S(t)$ ,  $0 \leq t \leq T_x$

## Goal:

- ▶ Calculate  $E[e^{-\delta T_x} \times \text{payment}]$
- ▶ the expectation of the discounted value of the payment

$\delta$  valuation force of interest

## Index process

- ▶  $S(t)$  price of one unit of a fund at time  $t$
- ▶  $S(t) = S(0)e^{X(t)}, \quad t \geq 0$
- ▶  $X(t)$  a Lévy process

## $\{X(t)\}$ a jump-diffusion process

$X(t) =$  Brownian motion ( $\mu, D = \sigma^2/2$ )

– compound Poisson ( $\omega, q(x)$ )  
downward jumps

+ compound Poisson ( $\nu, p(x)$ )  
upward jumps

## Double exponential jump-diffusion

- ▶  $\nu, p(x) = ve^{-vx}, x > 0$  upward
- ▶  $\omega, q(x) = we^{-wx}, x > 0$  downward
- ▶ a model with 6 parameters
- ▶ a reasonable compromise between Brownian motion and a general Lévy process

## More general jump-diffusions

- ▶  $\nu$ ,  $p(x) = \sum_{i=1}^n P_i v_i e^{-v_i x}$ ,  $x > 0$  upward
- ▶  $\omega$ ,  $q(x) = \sum_{i=1}^m Q_i w_i e^{-w_i x}$ ,  $x > 0$  downward
- ▶ Kou and his co-authors' work
- ▶ a reasonable compromise between Brownian motion and a general Lévy process

## $\Psi(z)$ : the Lévy exponent

- ▶  $E[e^{zX(t)}] = e^{t\Psi(z)}$
- ▶  $\Psi(z) = Dz^2 + \mu z - \nu \int_0^\infty (1 - e^{zx}) p(x) dx - \omega \int_0^\infty (1 - e^{-zx}) q(x) dx$
- ▶ For the double exponential jump-diffusion

$$\Psi(z) = Dz^2 + \mu z + \nu \frac{z}{v-z} - \omega \frac{z}{w+z}$$

- ▶  $\Psi(z) = 0$ : Lundberg's equation
- ▶  $\Psi(z) = \lambda$ : Generalized Lundberg's equation
- ▶ For the double exponential jump-diffusion, this equation has “four” solutions

$$-\infty < \alpha_2 < -w < \alpha_1 < 0 < \beta_1 < v < \beta_2 < \infty$$

## For Brownian motion

- ▶  $Dz^2 + \mu z = \lambda$ : Generalized Lundberg's equation
- ▶ Two solutions:  $\alpha < 0, \beta > 0$

## Connection with martingales

- ▶  $\{e^{-\lambda t} e^{zX(t)}\}$  is a martingale \*
- ↔  $z$  solution of the generalized Lundberg's equation
- ▶ Alternative formulation of \*  
 $\tau$  independent exponential random variable with parameter  $\lambda$ ,  
 $\Pr(\tau > t) = e^{-\lambda t}$   
 $\{I_{(\tau>t)} e^{zX(t)}\}$  martingale

## The reduced problem

- ▶ Idea: the pdf of  $T_x$  can be approximated by

$$\sum_{i=1}^n A_i \lambda_i e^{-\lambda_i t}, \quad t > 0$$

So, it suffices that we know how to calculate

$$E[e^{-\delta\tau} b(S(\tau))] = \int_0^\infty e^{-\delta t} \lambda e^{-\lambda t} E[b(S(t))] dt$$

## The reduced problem

$\delta$  can be eliminated

$$\begin{aligned}& \int_0^\infty e^{-\delta t} \lambda e^{-\lambda t} E[b(S(t))] dt \\&= \frac{\lambda}{\lambda + \delta} \int_0^\infty (\lambda + \delta) e^{-(\lambda + \delta)t} E[b(S(t))] dt\end{aligned}$$

rule: do the calculation without discounting  
but replace  $\lambda$  by  $\lambda + \delta$  multiply by  $\frac{\lambda}{\lambda + \delta}$

Note:

- ▶ If  $\lambda$  is replaced by  $\lambda + \delta$ ,

the generalized Lundberg's equation is

$$\Psi(z) = \lambda + \delta$$

- ▶ Thus the  $\alpha$ 's and  $\beta$ 's are modified accordingly

- ▶ Want to calculate

$$\mathbb{E}[e^{-\delta\tau} b(S(\tau))] = \mathbb{E}[e^{-\delta\tau} b(S(0)e^{X(\tau)})]$$

- ▶ so we need

$f_{X(\tau)}(x)$  the pdf of  $X(\tau)$

## Distribution of $X(\tau)$

$$\begin{aligned} \mathbb{E}[e^{zX(\tau)}] &= \mathbb{E}[\mathbb{E}[e^{zX(\tau)} \mid \tau]] = \mathbb{E}[e^{\tau\Psi(z)}] = \frac{\lambda}{\lambda - \Psi(z)} \\ &= \frac{\lambda}{\Psi'(\alpha_1)} \frac{1}{\alpha_1 - z} + \frac{\lambda}{\Psi'(\alpha_2)} \frac{1}{\alpha_2 - z} + \frac{\lambda}{\Psi'(\beta_1)} \frac{1}{\beta_1 - z} + \frac{\lambda}{\Psi'(\beta_2)} \frac{1}{\beta_2 - z} \\ \rightarrow f_{X(\tau)}(x) &= \begin{cases} a_1 e^{-\alpha_1 x} + a_2 e^{-\alpha_2 x}, & x < 0, \\ b_1 e^{-\beta_1 x} + b_2 e^{-\beta_2 x}, & x \geq 0 \end{cases} \end{aligned}$$

with

$$a_j = \frac{-\lambda}{\Psi'(\alpha_j)}, \quad b_j = \frac{\lambda}{\Psi'(\beta_j)}, \quad j = 1, 2.$$

$$b(s) = (K - s)_+ \text{ put option}$$

- ▶ out-of-the money:  $S(0) \geq K$

$$\mathcal{E}_b(S(0)) = E[(K - S(\tau))_+] = a_1 \eta(\alpha_1; K, S(0)) + a_2 \eta(\alpha_2; K, (S(0)))$$

with

$$\eta(h; K, S(0)) = \frac{K^{1-h} S(0)^h}{(h-1)h}$$

$b(s) = (s - K)_+$  call option

- ▶ out-of-the money:  $S(0) \leq K$

$$\mathcal{E}_b(S(0)) = b_1 \eta(\beta_1; K, S(0)) + b_2 \eta(\beta_2; K, S(0))$$

with

$$\eta(h; K, S(0)) = \frac{K^{1-h} S(0)^h}{(h-1)h}$$

## In-the-money formulas

Use put-call parity

$$[K - S(\tau)]_+ - [S(\tau) - K]_+ = K - S(\tau)$$

$$\mathbb{E}[[K - S(\tau)]_+] - \mathbb{E}[[S(\tau) - K]_+] = K - \mathbb{E}[S(\tau)],$$

## Running maximum

- ▶  $M(t) = \max\{X(u); 0 \leq u \leq t\}$  running maximum
- ▶ For the discussion of lookback call options,  
we need the distribution of  $M(\tau)$
- ▶ result:

$$f_{M(\tau)}(x) = \frac{\beta_2(v - \beta_1)}{v(\beta_2 - \beta_1)} \beta_1 e^{-\beta_1 x} + \frac{\beta_1(\beta_2 - v)}{v(\beta_2 - \beta_1)} \beta_2 e^{-\beta_2 x}, \quad x \geq 0,$$

$\beta_1$  and  $\beta_2$  are the positive solutions of  $\Psi(z) = \lambda$

## Proof

$$\Pr(M(\tau) \geq x) = \Pi_d(x) + \Pi_s(x),$$

where

$\Pi_d(x)$  is the probability that the process exceeds  $x$  before time  $\tau$  and when it occurs, it is because of oscillation.

$\Pi_s(x)$  is the probability that the process exceeds  $x$  before time  $\tau$  and when it occurs, it is because of an upward jump.

## Proof

Stop the martingale  $\{e^{\beta_1 X(t)} I_{(t < \tau)}; t \geq 0\}$  the first time when  $\{X(t)\}$  attains or jumps over level  $x$ . Optional sampling theorem yields

$$1 = e^{\beta_1 x} \Pi_d(x) + \frac{v}{v - \beta_1} e^{\beta_1 x} \Pi_s(x).$$

By analytical analogy we have

$$1 = e^{\beta_2 x} \Pi_d(x) + \frac{v}{v - \beta_2} e^{\beta_2 x} \Pi_s(x).$$

## Proof

Solution:

$$\begin{aligned}\Pi_d(x) &= \frac{(v - \beta_1)e^{-\beta_1 x} + (\beta_2 - v)e^{-\beta_2 x}}{\beta_2 - \beta_1}, \\ \Pi_s(x) &= \frac{(v - \beta_1)(\beta_2 - v)(e^{-\beta_1 x} - e^{-\beta_2 x})}{v(\beta_2 - \beta_1)}.\end{aligned}$$

Then

$$\Pr(M(\tau) \geq x) = \frac{\beta_2(v - \beta_1)e^{-\beta_1 x} + \beta_1(\beta_2 - v)e^{-\beta_2 x}}{v(\beta_2 - \beta_1)}, \quad x \geq 0,$$

$$f_{M(\tau)}(x) = \frac{\beta_2(v - \beta_1)}{v(\beta_2 - \beta_1)}\beta_1 e^{-\beta_1 x} + \frac{\beta_1(\beta_2 - v)}{v(\beta_2 - \beta_1)}\beta_2 e^{-\beta_2 x}, \quad x \geq 0,$$

## Running minimum

- ▶  $m(t) = \min\{X(u); 0 \leq u \leq t\}$  running minimum
- ▶ For the discussion of lookback put options,  
we need the distribution of  $m(\tau)$
- ▶ result:

$$f_{m(\tau)}(x) = \frac{-\alpha_2(\alpha_1 + w)}{w(\alpha_1 - \alpha_2)}(-\alpha_1 e^{-\alpha_1 x}) + \frac{\alpha_1(w + \alpha_2)}{w(\alpha_1 - \alpha_2)}(-\alpha_2 e^{-\alpha_2 x}),$$

$\alpha_1$  and  $\alpha_2$  are the negative solutions of  $\Psi(z) = \lambda$

# Proof

$$\min\{X(t); 0 \leq t \leq \tau\}$$

$$= -\max\{-X(t); 0 \leq t \leq \tau\}$$

## Lookback call options

$$b = [\max\{S(t); 0 \leq t \leq \tau\} - K]_+ = [S(0)e^{M(\tau)} - K]_+$$

out-of-the money:  $S(0) \leq K$

$$\begin{aligned} & \int_{\ln[K/S(0)]}^{\infty} [S(0)e^x - K]_+ f_{M(\tau)}(x) dx \\ &= \frac{\beta_1 \beta_2}{v(\beta_2 - \beta_1)} [(v - \beta_1)\eta(\beta_1; K) + (\beta_2 - v)\eta(\beta_2; K)] \end{aligned}$$

## Lookback put options

$$b = [K - \min\{S(t); 0 \leq t \leq \tau\}]_+ = [K - S(0)e^{m(\tau)}]_+$$

out-of-the money:  $S(0) \geq K$

$$\begin{aligned} & \int_{-\infty}^{\ln[S(0)/K]} [K - S(0)e^x] f_{m(\tau)}(x) dx \\ &= \frac{\alpha_1 \alpha_2}{w(\alpha_1 - \alpha_2)} [(\alpha_1 + w)\eta(\alpha_1; K) - (w + \alpha_2)\eta(\alpha_2; K)] \end{aligned}$$

## Joint pdf of $X(\tau)$ and $M(\tau)$

By

$$X(\tau) = M(\tau) - [M(\tau) - X(\tau)],$$

and independence of  $M(\tau)$  and  $M(\tau) - X(\tau)$ , the joint pdf of  $X(\tau)$  and  $M(\tau)$ , for  $y \geq \max(0, x)$ , is

$$\begin{aligned} f_{X(\tau), M(\tau)}(x, y) &= f_{M(\tau), M(\tau)-X(\tau)}(y, y-x) \\ &= f_{M(\tau)}(y) f_{M(\tau)-X(\tau)}(y-x) \\ &= \frac{\lambda}{D(\alpha_1 - \alpha_2)(\beta_2 - \beta_1)} \{-\}, \end{aligned}$$

with

$$\begin{aligned}\{-\} &= (\alpha_1 + w)(v - \beta_1)e^{-\alpha_1 x}e^{(\alpha_1 - \beta_1)y} \\ &\quad - (w + \alpha_2)(v - \beta_1)e^{-\alpha_2 x}e^{(\alpha_2 - \beta_1)y} \\ &\quad + (\alpha_1 + w)(\beta_2 - v)e^{-\alpha_1 x}e^{(\alpha_1 - \beta_2)y} \\ &\quad - (w + \alpha_2)(\beta_2 - v)e^{-\alpha_2 x}e^{(\alpha_2 - \beta_2)y}.\end{aligned}$$

## Single barrier options

- ▶ Up-and-in option,  $S(0) < L$ , notation:  $\ell = \ln[L/S(0)]$
- ▶ payoff at time  $\tau$

$$I_{(\max_{0 \leq t \leq \tau} S(t) \geq L)} b(S(\tau)) = I_{(M(\tau) \geq \ell)} b(S(0)e^{X(\tau)})$$

Expected payoff  $\int_{\ell}^{\infty} \left[ \int_{-\infty}^y b(S(0)e^x) f_{X(\tau), M(\tau)}(x, y) dx \right] dy$

Need the joint pdf of  $X(\tau)$  and  $M(\tau)$

## Alternative expression for expected payoff

$$\Pi_d(\ell) \mathcal{E}_b(L) + \Pi_s(\ell) \int_0^\infty \mathcal{E}_b(L e^x) v e^{-vx} dx,$$

where

$$\Pi_d(\ell) = \frac{(v - \beta_1)(\frac{S(0)}{L})^{\beta_1} + (\beta_2 - v)(\frac{S(0)}{L})^{\beta_2}}{\beta_2 - \beta_1},$$

$$\Pi_s(\ell) = \frac{(v - \beta_1)(\beta_2 - v)[(\frac{S(0)}{L})^{\beta_1} - (\frac{S(0)}{L})^{\beta_2}]}{v(\beta_2 - \beta_1)}.$$

For a particular payoff function  $b(s)$ , it remains to determine  $\mathcal{E}_b(L)$  and  $\int_0^\infty \mathcal{E}_b(L e^x) v e^{-vx} dx$ .

## Up-and-in put option, $L \geq K$

$$\mathcal{E}_b(L) = a_1\eta(\alpha_1; K, L) + a_2\eta(\alpha_2; K, L),$$

and

$$\int_0^\infty \mathcal{E}_b(L e^x) v e^{-vx} dx = \frac{v}{v - \alpha_1} a_1\eta(\alpha_1; K, L) + \frac{v}{v - \alpha_2} a_2\eta(\alpha_2; K, L),$$

where

$$\eta(h; K, L) = \frac{K^{1-h} L^h}{(h-1)h},$$

## Up-and-in put option, $L < K$

$$\mathcal{E}_b(L) = b_1 \eta(\beta_1; K, L) + b_2 \eta(\beta_2; K, L) + K - L \mathbb{E}[e^{X(\tau)}],$$

and

$$\begin{aligned} \int_0^\infty \mathcal{E}_b(L e^x) v e^{-vx} dx &= \frac{v}{v - \beta_1} \left[ 1 - \left( \frac{L}{K} \right)^{v - \beta_1} \right] b_1 \eta(\beta_1; K, L) \\ &\quad + \frac{v}{v - \beta_2} \left[ 1 - \left( \frac{L}{K} \right)^{v - \beta_2} \right] b_2 \eta(\beta_2; K, L) \\ &\quad + K - L^v K^{1-v} + L \frac{v}{1-v} \mathbb{E}[e^{X(\tau)}] \left[ 1 - \left( \frac{L}{K} \right)^{v-1} \right] \\ &\quad + \frac{v}{v - \alpha_1} a_1 \eta(\alpha_1; K, L) \left[ \frac{K}{L} \right]^{\alpha_1 - v} + \frac{v}{v - \alpha_2} a_2 \eta(\alpha_2; K, L) \left[ \frac{K}{L} \right]^{\alpha_2 - v}. \end{aligned}$$

## Brownian motion model

The expectation of the time- $\tau$  payoff of a up-and-in barrier option with barrier  $L$ ,  $L > S(0)$ , is

$$\left[ \frac{S(0)}{L} \right]^\beta \mathcal{E}_b(L)$$

For up-and-out case

$$E[I_{(M(\tau) < \ell)} b(S(\tau))] = \mathcal{E}_b(S(0)) - \left[ \frac{S(0)}{L} \right]^\beta \mathcal{E}_b(L).$$

The expectation of the time- $\tau$  payoff of a down-and-in barrier option with barrier  $L$ ,  $L > S(0)$ , is

$$\left[ \frac{L}{S(0)} \right]^{-\alpha} \mathcal{E}_b(L).$$

The up-and-out option formula can also be expressed as

$$\mathcal{E}_g(S(0)) - \left[ \frac{S(0)}{L} \right]^\beta \mathcal{E}_g(L),$$

where the function  $g$  is defined by

$$g(s) = I_{(s < L)} b(s),$$

and

$$\mathcal{E}_g(s) = E[g(S(\tau)) | S(0) = s] = E[I_{(S(\tau) < L)} b(S(\tau)) | S(0) = s].$$

Let  $s_2 > s_1 \geq L$ . If the initial stock price is  $s_2$ ,  $g(S(\tau)) = 0$  unless the stock price drops to the level  $s_1$  before time  $\tau$ , the probability of which is  $(s_1/s_2)^{-\alpha}$ .

$$\mathcal{E}_g(s_2) = \left( \frac{s_1}{s_2} \right)^{-\alpha} \mathcal{E}_g(s_1),$$

or

$$\mathcal{E}_g(s_1) = \left( \frac{s_1}{s_2} \right)^\alpha \mathcal{E}_g(s_2).$$

In particular,

$$\mathcal{E}_g(L) = \left[ \frac{S(0)}{L} \right]^\alpha \mathcal{E}_g\left( \frac{L^2}{S(0)} \right).$$

$$\begin{aligned}
 \mathbb{E}[I_{(M(\tau) < \ell)} b(S(\tau))] &= \mathcal{E}_g(S(0)) - \left[ \frac{S(0)}{L} \right]^{\alpha+\beta} \mathcal{E}_g\left( \frac{L^2}{S(0)} \right) \\
 &= \mathcal{E}_g(S(0)) - \left[ \frac{S(0)}{L} \right]^{-\mu/D} \mathcal{E}_g\left( \frac{L^2}{S(0)} \right).
 \end{aligned}$$

The result can be generalized to case where the barrier is an exponential function of time,  $Le^{\xi t}$ ,  $t \geq 0$ ,  $\xi$  being a real constant.

Similar results can be obtained for the double-barrier option.

## First exit from an interval

Let  $x_1 < x < x_2$  and let

$$\mathcal{T} = \min\{t : x + X(t) \leq x_1 \text{ or } x + X(t) \geq x_2\}$$

be the exit time of the process  $\{x + X(t)\}$  from the interval  $(x_1, x_2)$ . We are interested in

$$\Pi_1(x) = \Pr(x + X(\mathcal{T}) \leq x_1, \mathcal{T} < \tau),$$

$$\Pi_2(x) = \Pr(x + X(\mathcal{T}) \geq x_2, \mathcal{T} < \tau).$$

Results of this kind are needed in the analysis of double barrier options exercisable at time  $\tau$ .

## Four ways to exit the interval

$$\Pi_{1d}(x) = \Pr(x + X(\mathcal{T}) = x_1, \mathcal{T} < \tau),$$

$$\Pi_{1s}(x) = \Pr(x + X(\mathcal{T}) < x_1, \mathcal{T} < \tau),$$

$$\Pi_{2d}(x) = \Pr(x + X(\mathcal{T}) = x_2, \mathcal{T} < \tau),$$

$$\Pi_{2s}(x) = \Pr(x + X(\mathcal{T}) > x_2, \mathcal{T} < \tau).$$

Then

$$\Pi_1(x) = \Pi_{1d}(x) + \Pi_{1s}(x),$$

$$\Pi_2(x) = \Pi_{2d}(x) + \Pi_{2s}(x).$$

Use the martingales  $\{e^{\alpha_1(x+X(t))} I_{(t<\tau)}; t \geq 0\}$  and  $\{e^{\beta_1(x+X(t))} I_{(t<\tau)}; t \geq 0\}$ , combined with the optional sampling theorem and the memoryless property of the jump random variables, to see that

$$\Pi_{1d}(x)e^{\alpha_1 x_1} + \Pi_{1s}(x)e^{\alpha_1 x_1} \frac{w}{w + \alpha_1} + \Pi_{2d}(x)e^{\alpha_1 x_2} + \Pi_{2s}(x)e^{\alpha_1 x_2} \frac{v}{v - \alpha_1} \\ = e^{\alpha_1 x},$$

$$\Pi_{1d}(x)e^{\beta_1 x_1} + \Pi_{1s}(x)e^{\beta_1 x_1} \frac{w}{w + \beta_1} + \Pi_{2d}(x)e^{\beta_1 x_2} + \Pi_{2s}(x)e^{\beta_1 x_2} \frac{v}{v - \beta_1} \\ = e^{\beta_1 x}.$$

By analytical analogy we have

$$\Pi_{1d}(x)e^{\alpha_2 x_1} + \Pi_{1s}(x)e^{\alpha_2 x_1} \frac{w}{w + \alpha_2} + \Pi_{2d}(x)e^{\alpha_2 x_2} + \Pi_{2s}(x)e^{\alpha_2 x_2} \frac{v}{v - \alpha_2}$$

$$= e^{\alpha_2 x}$$

$$\Pi_{1d}(x)e^{\beta_2 x_1} + \Pi_{1s}(x)e^{\beta_2 x_1} \frac{w}{w + \beta_2} + \Pi_{2d}(x)e^{\beta_2 x_2} + \Pi_{2s}(x)e^{\beta_2 x_2} \frac{v}{v - \beta_2}$$

$$= e^{\beta_2 x}.$$

- ▶ Use Cramer's rule to obtain  $\Pi_{1d}(x)$ ,  $\Pi_{1s}(x)$ ,  $\Pi_{2d}(x)$ ,  $\Pi_{2s}(x)$
- ▶ Each is a linear combination of  $e^{\alpha_1 x}$ ,  $e^{\alpha_2 x}$ ,  $e^{\beta_1 x}$ ,  $e^{\beta_2 x}$

## Solutions of integro-differential equations

$$\Pi_{1d}(x) : \quad \mathcal{L}\phi(x) = 0, \quad x_1 \leq x \leq x_2, \quad \phi(x_1) = 1, \quad \phi(x_2) = 0$$

$$\Pi_{2d}(x) : \quad \mathcal{L}\phi(x) = 0, \quad x_1 \leq x \leq x_2, \quad \phi(x_1) = 0, \quad \phi(x_2) = 1$$

$$\Pi_{1s}(x) : \quad \mathcal{L}\phi(x) + \omega e^{-w(x-x_1)} = 0, \quad \phi(x_1) = \phi(x_2) = 0$$

$$\Pi_{2s}(x) : \quad \mathcal{L}\phi(x) + \nu e^{-v(x_2-x)} = 0, \quad \phi(x_1) = \phi(x_2) = 0$$

with

$$\begin{aligned} \mathcal{L}\phi(x) &= D\phi''(x) + \mu\phi'(x) - (\lambda + \nu + \omega)\phi(x) \\ &\quad + \nu v \int_0^{x_2-x} \phi(x+y)e^{-vy} dy + \omega w \int_0^{x-x_1} \phi(x-y)e^{-wy} dy \end{aligned}$$

## Substitute

$$\phi(x) = A_1 e^{\alpha_1 x} + A_2 e^{\alpha_2 x} + B_1 e^{\beta_1 x} + B_2 e^{\beta_2 x}.$$

and get each time a system of four linear equations:

$$\begin{aligned} A_1 e^{\alpha_1 x_2} \frac{v}{v - \alpha_1} + A_2 e^{\alpha_2 x_2} \frac{v}{v - \alpha_2} + B_1 e^{\beta_1 x_2} \frac{v}{v - \beta_1} + B_2 e^{\beta_2 x_2} \frac{v}{v - \beta_2} &= 0 \\ A_1 e^{\alpha_1 x_1} \frac{w}{w + \alpha_1} + A_2 e^{\alpha_2 x_1} \frac{w}{w + \alpha_2} + B_1 e^{\beta_1 x_1} \frac{w}{w + \beta_1} + B_2 e^{\beta_2 x_1} \frac{w}{w + \beta_2} &= 0 \\ A_1 e^{\alpha_1 x_1} + A_2 e^{\alpha_2 x_1} + B_1 e^{\beta_1 x_1} + B_2 e^{\beta_2 x_1} &= 1 \\ A_1 e^{\alpha_1 x_2} + A_2 e^{\alpha_2 x_2} + B_1 e^{\beta_1 x_2} + B_2 e^{\beta_2 x_2} &= 0 \end{aligned}$$

for  $\phi(x) = \Pi_{1d}(x)$ .

## Substitute

$$\phi(x) = A_1 e^{\alpha_1 x} + A_2 e^{\alpha_2 x} + B_1 e^{\beta_1 x} + B_2 e^{\beta_2 x}.$$

and get each time a system of four linear equations:

$$\begin{aligned} A_1 e^{\alpha_1 x_2} \frac{v}{v - \alpha_1} + A_2 e^{\alpha_2 x_2} \frac{v}{v - \alpha_2} + B_1 e^{\beta_1 x_2} \frac{v}{v - \beta_1} + B_2 e^{\beta_2 x_2} \frac{v}{v - \beta_2} &= 0 \\ A_1 e^{\alpha_1 x_1} \frac{w}{w + \alpha_1} + A_2 e^{\alpha_2 x_1} \frac{w}{w + \alpha_2} + B_1 e^{\beta_1 x_1} \frac{w}{w + \beta_1} + B_2 e^{\beta_2 x_1} \frac{w}{w + \beta_2} &= 0 \\ A_1 e^{\alpha_1 x_1} + A_2 e^{\alpha_2 x_1} + B_1 e^{\beta_1 x_1} + B_2 e^{\beta_2 x_1} &= 0 \\ A_1 e^{\alpha_1 x_2} + A_2 e^{\alpha_2 x_2} + B_1 e^{\beta_1 x_2} + B_2 e^{\beta_2 x_2} &= 1 \end{aligned}$$

for  $\phi(x) = \Pi_{2d}(x)$ .

## Substitute

$$\phi(x) = A_1 e^{\alpha_1 x} + A_2 e^{\alpha_2 x} + B_1 e^{\beta_1 x} + B_2 e^{\beta_2 x}.$$

and get each time a system of four linear equations:

$$\begin{aligned} A_1 e^{\alpha_1 x_2} \frac{v}{v - \alpha_1} + A_2 e^{\alpha_2 x_2} \frac{v}{v - \alpha_2} + B_1 e^{\beta_1 x_2} \frac{v}{v - \beta_1} + B_2 e^{\beta_2 x_2} \frac{v}{v - \beta_2} &= 0 \\ A_1 e^{\alpha_1 x_1} \frac{w}{w + \alpha_1} + A_2 e^{\alpha_2 x_1} \frac{w}{w + \alpha_2} + B_1 e^{\beta_1 x_1} \frac{w}{w + \beta_1} + B_2 e^{\beta_2 x_1} \frac{w}{w + \beta_2} &= 1 \\ A_1 e^{\alpha_1 x_1} + A_2 e^{\alpha_2 x_1} + B_1 e^{\beta_1 x_1} + B_2 e^{\beta_2 x_1} &= 0 \\ A_1 e^{\alpha_1 x_2} + A_2 e^{\alpha_2 x_2} + B_1 e^{\beta_1 x_2} + B_2 e^{\beta_2 x_2} &= 0 \end{aligned}$$

for  $\phi(x) = \Pi_{1s}(x)$ .

## Substitute

$$\phi(x) = A_1 e^{\alpha_1 x} + A_2 e^{\alpha_2 x} + B_1 e^{\beta_1 x} + B_2 e^{\beta_2 x}.$$

and get each time a system of four linear equations:

$$\begin{aligned} A_1 e^{\alpha_1 x_2} \frac{v}{v - \alpha_1} + A_2 e^{\alpha_2 x_2} \frac{v}{v - \alpha_2} + B_1 e^{\beta_1 x_2} \frac{v}{v - \beta_1} + B_2 e^{\beta_2 x_2} \frac{v}{v - \beta_2} &= 1 \\ A_1 e^{\alpha_1 x_1} \frac{w}{w + \alpha_1} + A_2 e^{\alpha_2 x_1} \frac{w}{w + \alpha_2} + B_1 e^{\beta_1 x_1} \frac{w}{w + \beta_1} + B_2 e^{\beta_2 x_1} \frac{w}{w + \beta_2} &= 0 \\ A_1 e^{\alpha_1 x_1} + A_2 e^{\alpha_2 x_1} + B_1 e^{\beta_1 x_1} + B_2 e^{\beta_2 x_1} &= 0 \\ A_1 e^{\alpha_1 x_2} + A_2 e^{\alpha_2 x_2} + B_1 e^{\beta_1 x_2} + B_2 e^{\beta_2 x_2} &= 0 \end{aligned}$$

for  $\phi(x) = \Pi_{2s}(x)$ .

## Double barrier knock-in option

- ▶  $0 < L < S(0) < U$
- ▶ barrier levels  $L, U$ ; initial stock price  $S(0)$
- ▶ The option comes into existence if one of the two barriers is reached before time  $\tau$
- ▶ Payoff  $I_{(S(0)e^{m(\tau)} \leq L, \text{ or } S(0)e^{M(\tau)} \geq U)} b(S(\tau))$

- ▶ Expected payoff of the double-barrier knock-in option is

$$\begin{aligned} & \Pi_{1d}(\ln S(0))\mathcal{E}_b(L) + \Pi_{1s}(\ln S(0)) \int_0^{\infty} \mathcal{E}_b(L e^{-x}) w e^{-wx} dx \\ & + \Pi_{2d}(\ln S(0))\mathcal{E}_b(U) + \Pi_{2s}(\ln S(0)) \int_0^{\infty} \mathcal{E}_b(U e^x) v e^{-vx} dx, \end{aligned}$$

with  $x_1 = \ln \frac{L}{S(0)}$  and  $x_2 = \ln \frac{U}{S(0)}$ .