

$\tilde{I}_{j+1,i}$: This part must remain explicit, otherwise it is very hard to deal with.

$j+1$ KNOWN ,

j UNKNOWN .

Now, let's implement this : we'll start from the Q-method that we implemented for BS & we'll modify it for Levy.

- We need new market parameters : $\delta, \underbrace{\min, \text{delta}}_{\text{jumptimes}}, \underbrace{\text{jumpsite} \sim N(\mu, \sigma^2)}$.
- Maybe enlarge the grid bounds : since BS underestimates large events ... You can enlarge it by starting from the BS bounds $(r - \frac{\sigma^2}{2})T \pm 6\sqrt{T}\sigma$; or by rule of thumbs, like

$$\left\{ x_{\min} = \log \left(\frac{0.2 s_0}{s_0} \right) = \log(0.2) ; \right.$$

$$\left. x_{\max} = \log \left(\frac{3 s_0}{s_0} \right) = \log(3) . \right.$$

• Before dealing with matrices, we must compute α & λ :

1) code a ν fct:

$\text{nu}@\{y\} \rightarrow \text{Lévy Measure}$.

2) code a fct:

$[\alpha, \lambda, \text{num}, lb, ub] = \text{Lévy_integral}(\nu)$

where we compute α & λ .

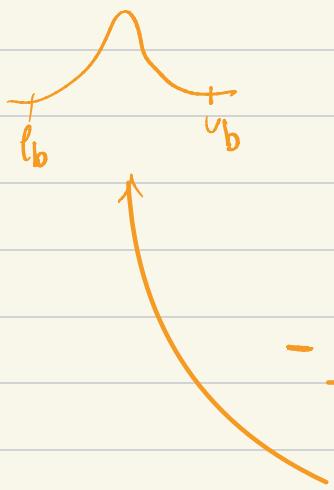
- Integral domain truncation:

we set our lb as the y s.t $\nu(y) < 10^{-10}$ on the left (so any point smaller than lb is $< 10^{-10}$).
Same for the upper bound.

- Quadrature: $N_q = 2N$;

we can plot the "kernel", i.e.,
the Lévy measure between lb & ub .

→ use for example "trapz()" for
trapezoidal quadrature formula.



3) change the matrices :

everywhere we have $r - \frac{\sigma^2}{2}$ we must add $(-\alpha)$ & everywhere r we must add λ .

4) In backward in time solution, we must add the integral I at time t_j . Like this:

let's write what is known on the right hand side & what is unknown on the lhs:

$$\begin{aligned} & -\frac{1}{\Delta t} V_{j,i} + (1-\theta) \left[\left(r - \frac{\sigma^2}{2} - \alpha \right) \frac{V_{j,i+1} - V_{j,i-1}}{2\Delta x} + \frac{\sigma^2}{2} \frac{V_{j,i+1} - 2V_{j,i} + V_{j,i-1}}{\Delta x^2} - \right. \\ & \quad \left. (r + \lambda) V_{j,i} \right] \\ & = -\frac{1}{\Delta t} V_{j+1,i} - \theta \left[\left(r - \frac{\sigma^2}{2} - \alpha \right) \frac{V_{j+1,i+1} - V_{j+1,i-1}}{2\Delta x} + \frac{\sigma^2}{2} \frac{V_{j+1,i+1} - 2V_{j+1,i} + V_{j+1,i-1}}{\Delta x^2} \right. \\ & \quad \left. - (r + \lambda) V_{j+1,i} \right] - \tilde{I}_{j+1,i} \end{aligned}$$

So this becomes:

$$M_1 v_j = M_2 v_{j+1} + BC - I$$

At each time step, we need to re-compute the integral I : $\text{Lvy_Integral2}(lb, ub, x, V, w)$.

Another fact with unique output I & computes the I integral.

We need an interpolation function to compute $v(t, aty)$, knowing V and x ...

⚠ when we change the BCs, we also have to change BCs in the integral's definition.

This is the main changing point w.r.t B&S PDE.

→ For Barrier Options, the barrier gives one of the bounds x_{\min} or x_{\max} . For example, for Down & Out, $x_{\min} = \log(\frac{D}{S_0})$.

→ For AM Int Option : don't use the discount factor on the BC.

And solve the linear system by PSOR to discriminate - step by step - whether we

are in Early Exercise or Continuation.

PS: we can do the same for Kou.

So what is missing here is to know how to deal with the Lévy processes that DON'T belong to F.A. Lévy \rightarrow I.A Lévy.

Recap: PIDE (r, σ^2, v) $\xrightarrow{\text{RN r.i.e}} \psi_x(-i) = 0$ (x fixed).

$S_t = S_0 e^x$; $f(t, x) \xleftarrow{\text{(EU) Call Option}} c(T, x) = (S_0 e^x - K)^+$

$$\underbrace{\frac{\partial c}{\partial t} + \left(r - \frac{\sigma^2}{2}\right) \frac{\partial c}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2 c}{\partial x^2} - rf}_{\text{BFS}} + \int_{\mathbb{R}} \left(c(t, x+y) - c(t, x) - (e^y - 1) \frac{\partial c}{\partial x} \right) \nu(y) dy$$

where $c := c(t, x)$.

1st IDEA: F.A so we can split the integral.

$$d = \int_{\mathbb{R}} (e^y - 1) \nu(y) dy, \quad \lambda = \int_{\mathbb{R}} \nu(y) dy \quad \text{then:}$$

$$\frac{\partial c}{\partial t} + \left(r - \frac{\sigma^2}{2} - \alpha\right) \frac{\partial c}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2 c}{\partial x^2} - (r + \lambda) c + \int_{\mathbb{R}} c(t, x+y) \nu(y) dy$$

This part is discretized as we want:

- E.E ;

- I.E ;

- Crank-Nicholson ...

→ Unconditionally stable.

only EXPLICITELY.



We know how to do it since it's exactly like B&S, but w/ α, λ .

• Let's discretize in time :

$$c(t, x) \quad t_j = j \Delta t \quad \text{so we have :}$$

$$\begin{aligned} \frac{1}{2}\text{-discretization: } & \left\{ \frac{c(t_{j+1}, x) - c(t_j, x)}{\Delta t} + \left(r - \frac{\sigma^2}{2} - \alpha\right) \frac{\partial c}{\partial x}(t_j, x) + \right. \\ & \left. \frac{\sigma^2}{2} \frac{\partial^2 c}{\partial x^2}(t_j, x) - (r + \lambda) c(t_j, x) + \int_R c(t_{j+1}, x+y) v(y) dy \right\} = 0 \end{aligned}$$

Now let's put the unknown on the left, the known on $J(t_{j+1}, x)$ the rhs :

$$-\frac{1}{\Delta t} c(t_j, x) + \left(r - \frac{\sigma^2}{2} - \alpha\right) \frac{\partial c}{\partial x}(t_j, x) + \frac{\sigma^2}{2} \frac{\partial^2 c}{\partial x^2}(t_j, x) - (r + \lambda) c(t_j, x)$$

$$= -\frac{1}{\Delta t} f(t_{j+1}, x) - J(t_{j+1}, x)$$

• let's discretize also in x (by pure):

$$x_i, i \in [0, N] . \quad \begin{cases} x_i = x_{\min} + i \Delta x \\ \text{where } \Delta x = \frac{x_{\max} - x_{\min}}{N} \end{cases}$$

BCs. $\begin{cases} c(t_j, x_0) = 0 \quad \forall j \\ c(t_j, x_N) = S_0 e^{x_N} - K e^{-r(T-t_j)} \quad \forall j \end{cases}$

And inside the domain ($i \in [1, N-1]$):

$$-\frac{1}{\Delta t} c_{j,i} + \left(r - \frac{\sigma^2}{2} - \alpha \right) \frac{c_{j,i+1} - c_{j,i-1}}{2 \Delta x} + \frac{\sigma^2}{2} \frac{c_{j,i+1} - 2c_{j,i} + c_{j,i-1}}{\Delta x^2}$$

$$-(r + \lambda) c_{j,i} = -\frac{1}{\Delta t} c_{j+1,i} - J(t_{j+1}, x_i) .$$

Explicit.



For a Put, in the code we need to change the payoff, the BCs **BUT ALSO** the Lévy Integral. The difference is in the BC, when $z <= x(1)$.



Possible to check with CM using P-C parity.