2021 4213002 京丹怡

1. 证明:

 $\kappa_{\lambda} = \lambda k_1 + (1-\lambda) k_2 > \lambda k_2 + (1-\lambda) k_2 = k_2$ 

記 Φ1: Pt(kλ) Φ2: λ Pt(k1) + (1-λ) Pt(k2). 苦 τ1=T. 则有

若 T1 = T. 则有

VT, (\$1) = PT (Kx) = (Kx-ST)+

 $V_{T_1}(\Phi_2) = \Lambda P_T(k_1) + (1-\lambda)P_T(k_2) = \lambda(k_1 - S_T)^* + (1-\lambda)(k_2 - S_T)^+$ 

n x is out t

太大海风采和 1(里)多0

$$= \lambda P_{T}(K_{1}) + (1-\lambda L)P_{T}(K_{2}) - \lambda L(K_{1} - S_{1}) + (1-\lambda L)P_{T}(K_{2}) + (1-\lambda$$

3 VT1(\$1) - 14 18 18 (4.14)

若 T, CT, 由美式看 跌期性质有

 $V_{T_i}(\Phi_i) = P_{T_i}(K_{\Lambda}) = (K_{\Lambda} - S_{T_i})^{\dagger}$ 

 $V_{T_1}(\underline{\Phi}_1) = \lambda P_{T_1}(k_1) + (1-\lambda)P_{T_1}(k_2)$ 

 $= \lambda (k_1 - S_{\tau_1}) + (1 - \lambda) (k_2 - S_{\tau_1})^{+}$ 

 $= \begin{cases} k_{\lambda} - S_{\tau_{1}} & (S_{\tau_{1}} - K_{2}) \\ \lambda(k_{1} - S_{\tau_{1}}) & (k_{2} < S_{\tau_{1}} - K_{1}) \end{cases}$ 

 $= \begin{cases} k_{\lambda} - ST_1, & ST_1 \leq k_2 \end{cases}$ 

( KA - SCI + (1-2)(SCI - K2) , K2 = SCI < K1

7 VT, (1)

综上所述,始终有 VT1(更2) > VT1(更1),由无套利原理 元有 Vt(平2) ≥ Vt(至1)

Pt (K2) = 2 Pt (K1) + (1-2) Pt (K2), t = T1 < T. a.s. =

若 $T_1 = T$ . 则有  $V_T(\Phi_1) = (K_1 - S_T)^+ + K_2 e^{\gamma(T-t)}$   $V_T(\Phi_2) = (K_2 - S_T)^+ + K_1 e^{\gamma(T-t)}$ 

 $|P| V_{T}(\Phi_{2}) = (k_{1}-k_{2})e^{Y(T-t)} + (k_{3}-S_{T})^{T} - (k_{1}-S_{T})^{T}$   $= \begin{cases} (k_{1}-k_{2})e^{Y(T-t)} - (k_{1}-k_{2}) & , S_{1} \geq k_{2} \leq k_{1} \\ (k_{1}-k_{2})e^{Y(T-t)} & -k_{1}+S_{T} & k_{2} \leq S_{T} \leq k_{1} \\ (k_{1}-k_{2})e^{Y(T-t)} & , S_{T} \geq k_{1} \geq k_{2} \end{cases}$   $= \begin{cases} (k_{1}-k_{2})e^{Y(T-t)} & , S_{T} \geq k_{1} \geq k_{2} \end{cases}$   $= \begin{cases} (k_{1}-k_{2})e^{Y(T-t)} & , S_{T} \geq k_{1} \geq k_{2} \end{cases}$ 

缘上所进由无套利原理 Vt(里,)≤ Vt(里2)

. ②证左边式子. \*(3(24),1)(1/2) \*\*(130-14)(15

記車: Pt(K1) - Pt(K2). (2) 10 (10) +

若  $\tau_1 = \tau$ . 则  $V_{\tau}(\Phi) = (k_1 - S_{\tau})^{+} - (k_2 - S_{\tau})^{+}$   $= 5 k_1 - k_2 \qquad 5 k_2 < k_1$   $= 6 k_1 - 5 k_2 \qquad k_2 < 5 k_1$   $= 6 k_1 - 5 k_2 \qquad k_3 < k_1$ 

若  $T_1 < T$  . 即  $V_{T_1}(\Phi) = k_1 - S_{T_1} - (k_2 - S_{T_2})^+$   $= \begin{cases} k_1 - k_2 & , S_{T_1} < k_2 < k_1 \end{cases}$   $= \begin{cases} k_1 - S_{T_1} & , k_2 < k_1 < k_2 < k_1 \end{cases}$ 

故由无套利 V(重)≥0

⇒ 0 < Pt(K1) - Pt(K2) < K1 - K2 , t < T1 ≤ T - a.S.

2. 解:

原生真是可怕忘知如此之川是明即有中已根壁。 α〉 t 时刻持有 Pt , 找 Δ 份 St 对冲风险 即 t 时刻组合为 Pt - Δ St, 在 t+at 用刻 则有 (168:11 = 168-KD = 11:89K

 $\bar{P}_{t}^{u} - \Delta S_{t}u = \bar{P}_{t}^{d} - \Delta S_{t}d = (\bar{P}_{t} - \Delta S_{t})P$ 

其中Stu为原生资产St的上升价格, Std为飞路价格, P=1+YAt

 $0 - \Delta \cdot S_{t}u = 1 - \Delta \cdot S_{t}d = (\vec{p}_{t} - \Delta S_{t}) \cdot P$ 

 $\Rightarrow \Delta = \frac{0-1}{(u-d)S_t} = -\frac{1}{(u-d)S_t}$ 

 $\bar{P}_{t} = \frac{1}{\ell} \times \frac{u - \rho}{u - d}$ 

1 (K-St) 1 KISST b>  $\overline{C_t + P_t}$  是一个无风险 投资组合, $\overline{C_{t+\Delta t} + P_{t+\Delta t}} = 1 \Rightarrow \overline{C_t + P_b} = \frac{1}{P}$  $\overline{C}_{c} = \frac{1}{e} - \overline{P}_{o} = \frac{1}{e} \cdot \frac{e - d}{u - d}$  1 = e = 0 1 = 0 1 =

MIN St = Stu\*Ct + Std\*Pt

3. 解:

 $dv = \left\{ \frac{\partial V}{\partial t} + \frac{1}{2} \left[ \frac{\sigma_1^2}{S_1} \frac{\partial^2 V}{\partial S_1^2} + 2 P \rho_1 \Gamma_2 S_1 S_2 \frac{\partial^2 V}{\partial S_1 \partial S} + \frac{\partial^2 S_2^2}{\partial S_2^2} \right] \right\} dt$ 

 $+\frac{\partial v}{\partial s_1} ds_1 + \frac{\partial v}{\partial s_2} ds_2$ 

 $dV = \frac{\partial V}{\partial t} + \frac{n}{|z|} \frac{\partial V}{\partial S_i} dS_i + \frac{1}{2} \frac{\partial}{|z|} \nabla_i^2 S_i^2 \frac{\partial^2 V}{\partial S_i^2}$ 

+ Rije Pij Vi Oj Si Sj DSi DSj A THE TOTAL I XAME TO THE TENER OF THE PARTY OF THE

The d= Try - (1-6 + E) 24 = P 44

7 = 5.40 + 6.6.6 - 2:) AC

: 4545 - 1 (4.4) = 1 1843 H3

1857) = (5-k) , 0 = t = [

4. 解:

PDE 模型:

$$\begin{cases} \frac{\partial V}{\partial t} + \frac{1}{3} \sigma^2 s^2 \frac{\partial^2 V}{\partial s^2} + (\gamma - Q) s \frac{\partial V}{\partial s} - \gamma V = 0 \\ V(S,T) = 0, & K_1 < S < K_2 \end{cases}$$

$$V(k_1,t) = V_p(k,t) , o < t < T$$

$$V(k_2,t) = V_c(k,t) , o < t < T$$
概率模型.

概弄模型

$$V(s,t) = e^{-rT} E^{\alpha} [f(s_T) | S_t = s]$$

$$f(s_T) = \begin{cases} (s_T - k)^+, & s_T > k_2 \\ 0, & k_1 < s_T < k_2 \end{cases}$$

$$| (k - s_T)^+, & k_2 < s_T$$

5. 解:

$$\begin{cases} \min \left\{ -LV \right\}, V - (S - K)^{\dagger} \right\} = 0, \quad 0 \leq S \leq L, \quad 0 \leq t \leq T \\ V(L,t) = L - K, \quad 0 \leq t \leq T \\ V(S,T) = (S - K)^{\dagger}, \quad 0 \leq t \leq T \end{cases}$$

$$\frac{1}{4^{\dagger}} -LV = -\left[\frac{\partial V}{\partial t} + \frac{T^{2}}{2}S^{2}\frac{\partial^{2}V}{\partial S^{2}} + (T - Q)S\frac{\partial V}{\partial S} - \gamma V\right]$$

引入惩罚函数  $F_{\varepsilon}(V^{\varepsilon}-g) = -\frac{1}{\varepsilon}(V^{\varepsilon}-g)$  ,  $p = (S-k)^{\dagger}$  ,  $\varepsilon > 0$ 刚 tm→ t"有  $-8\frac{1}{5}V_{i}^{n+1} - \frac{\sigma^{2}}{2}8^{2}_{x}V_{i}^{n+1} - (\gamma - 9 - \frac{\sigma^{2}}{2})8xV_{i}^{n+1} + \gamma V_{i}^{n} + F_{e}(V_{i}^{n+1} - 9) = 0$ 

$$\frac{1}{4} V_{i}^{n+1} - \frac{1}{2} 8x^{2} V_{i}^{n} - (\Upsilon - Q - Z) \delta x V_{i} + V_{i} + F_{8}(V_{i}^{n} - Q) = V_{j}^{n} = \max \left\{ \frac{1}{1 + r_{At}} \left[ c_{A} V_{j-1}^{n+1} + c_{A} V_{j+1}^{n+1} \right], g_{i} \right\}$$

$$V_{j}^{N} = g_{j}$$
 ,  $V_{M}^{n} = L - K$ 

$$\frac{1}{2} \frac{dr}{dr} = \frac{\sigma^2 At}{2(AX)^2} - (r - q - \frac{\sigma^2}{2}) \frac{\Delta t}{2AX} \qquad \beta = 1 - \frac{\sigma^2 At}{(AX)^2}$$

$$\gamma = \frac{\sigma^2 At}{2(AX)^2} + (\gamma - q - \frac{\sigma^2}{2}) \frac{\Delta t}{2AX}$$

6. 解:

在 
$$i-M < x < M$$
 ,  $-M < y < M$  ,  $0 \le t \le T$   $j \ge 2$   $j \le M$   $j = j \triangle y$  ,  $-\frac{M}{\Delta y} < j < \frac{M}{\Delta y}$ 

$$V|_{x=-M}=e^y$$
  $V|_{y=-M}=e^x$   $V|_{x=M}=e^M$   $V|_{y=M}=e^M$ 

$$\frac{i\mathcal{C} V_{ij}^{n} = V(\chi_{i}, y_{j}, t_{n})}{2t} = \frac{V_{ij}^{n+1} - V_{ij}^{n}}{2t} = \frac{\partial V}{\partial \chi^{2}} = \frac{V_{in,j}^{n+1} - V_{in,j}^{n+1}}{2\Delta \chi} = \frac{\partial V}{\partial \chi^{2}} = \frac{V_{in,j}^{n+1} - V_{in,j}^{n+1}}{2\Delta \chi} = \frac{\partial V}{\partial \chi^{2}} = \frac{V_{in,j}^{n+1} - V_{in,j}^{n+1}}{2\Delta \chi} = \frac{\partial^{2} V}{\partial \chi^{2}} = \frac{V_{in,j}^{n+1} - 2V_{ij}^{n+1} - 2V_{ij}^{n+1} - 2V_{ij}^{n+1}}{(\Delta \chi)^{2}}$$

代入司求得  

$$V_{ij}^{n} = \frac{1}{1+T\Delta t} \left[ \left( 1 - \frac{\sigma_{i}^{2}\Delta t}{(\Delta \chi)^{2}} - \frac{\sigma_{2}^{2}\Delta t}{(\Delta y)^{2}} - e^{\sigma_{i}\sigma_{2}} \cdot \frac{\Delta t}{\Delta \chi_{\Delta y}} \right) V_{ij}^{n+1} \right]$$

$$+\left(\frac{\overline{v_{i}}^{2}\Delta t}{2(\Delta X)^{2}}+\frac{\Delta t}{2\Delta X}\left(\gamma-\frac{\overline{v_{i}}^{2}}{2}\right)+\frac{\rho \overline{v_{i}} \overline{v_{2}}\Delta t}{2\Delta X \Delta y}\right)V_{i+i,j}^{n+i}$$

$$+\left(\frac{\overline{v_{i}}^{2}\Delta t}{2(\Delta X)^{2}}-\frac{\Delta t}{2\Delta X}\left(\gamma-\frac{\overline{v_{i}}^{2}}{2}\right)+\frac{\rho \overline{v_{i}} \overline{v_{2}}\Delta t}{2\Delta X \Delta y}\right)V_{i-i,j}^{n+i}$$

+ 
$$\left(\frac{\overline{U_{2}^{2}}At}{2(\Delta y)^{2}} + \frac{At}{2\Delta y}(\gamma - \frac{\overline{U_{2}^{2}}}{2}) + \frac{\rho \overline{U_{1}} \overline{U_{2}} \Delta t}{2\Delta \chi \Delta y}\right) V_{i,f+1}^{n+1}$$

$$+\left(\frac{\sigma_{2}^{2}At}{2Ay}\right)^{2}-\frac{At}{2Ay}\left(\gamma-\frac{\sigma_{3}^{2}}{2}\right)+\frac{\varrho\sigma_{i}\sigma_{2}At}{2A\chi Ay}\left(\gamma-\frac{\eta+1}{2}\right)$$

$$= \frac{1}{1+rat} \left[ \alpha V_{ij}^{n+1} + \beta_{+} V_{i+1,j}^{n+1} + \beta_{-} V_{i+1,j}^{n+1} + \gamma_{+} V_{i,j+1}^{n+1} + \gamma_{-} V_{i,j-1}^{n+1} + \gamma_{+} V_{i-1,j+1}^{n+1} + \gamma_{-} V_{i+1,j-1}^{n+1} \right]$$

$$i \mathcal{C} d_1 = \frac{\sigma^2 \Delta t}{(\Delta X)^2} > 0 \quad d_2 = \frac{\sigma_2^2 \Delta t}{(\Delta Y)^2} > 0. \quad \mathcal{U} = \mathcal{U} + = \mathcal{U} - \frac{\sigma_2^2 \Delta t}{(\Delta Y)^2} > 0.$$

该柏式保持稳.定性.要求 d, β±, γ±, μ± ≥0

则需要满足以下不等式组

$$\begin{cases}
d = 1 - \alpha_{1} - d_{2} + 2\mu \geqslant 0 & 0 \\
\theta + = \frac{\alpha_{1}}{2} \left[ 1 + \frac{1}{\sigma_{1}^{2}} (r - \frac{\sigma_{1}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
\theta - = \frac{\alpha_{1}}{2} \left[ 1 - \frac{1}{\sigma_{1}^{2}} (r - \frac{\sigma_{1}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
\gamma + = \frac{\alpha_{1}}{2} \left[ 1 + \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
\gamma - = \frac{\alpha_{2}^{2}}{2} \left[ 1 - \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
0 - \frac{\alpha_{1}^{2}}{2} \left[ 1 - \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
0 - \frac{\alpha_{2}^{2}}{2} \left[ 1 - \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
0 - \frac{\alpha_{2}^{2}}{2} \left[ 1 - \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
0 - \frac{\alpha_{2}^{2}}{2} \left[ 1 - \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
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0 - \frac{\alpha_{2}^{2}}{2} \left[ 1 - \frac{1}{\sigma_{2}^{2}} (r - \frac{\sigma_{2}^{2}}{2}) \Delta x \right] - \mu \geqslant 0 & 0 \\
0 - \frac{\alpha_{2}^{2}}{2} \left[ 1 - \frac{1}$$

助 P<0, 5, 52, At, AX, Ay > 0 故 ル+=ル->0 恒成立 由①可得 d,+ d2 ≤ 1 + Z/L 由 ② + ② 可得 以 3 2 从 由 ④ + ⑤ 可得 对 2 2 2 从 故有 ZM = d1 = 1 2M = d2 = 1 使 ②③④⑤ 成立 还要有 自然意义中国第二十五 17 + Vija + Viij + Vija

当[1一点 17一型1001-1120 2 [1- 12-17- 12-1AY]-120 100 = 100 (00 - 100) - 100 - 100)

⇒ 
$$\frac{1}{\sigma_{1}^{2}}\left(1\gamma-\frac{\sigma_{1}^{2}}{2}1-\frac{\rho\sigma_{1}}{\Delta y}\right)\Delta\chi\leq 1$$
  
 $\frac{1}{\sigma_{2}^{2}}\left(1\gamma-\frac{\sigma_{2}^{2}}{2}1-\frac{\rho\sigma_{2}}{\Delta x}\right)Ay\leq 1$   
故使  $d$ ,  $\beta_{\pm}$ ,  $\gamma_{\pm}$ ,  $M_{\pm}\geq 0$  閉条件为

数使 d, 
$$\beta$$
± ,  $\gamma$ ± ,  $\mu$ ±  $\geq 0$  闭条件为
$$\begin{cases}
-\frac{\rho\sigma_2}{\sigma_1} \leq \frac{\Delta y}{\Delta x}, & -\frac{\rho\sigma_1}{\sigma_2} \leq \frac{\Delta x}{\Delta y} \\
\frac{\sigma_1^2 \Delta t}{(\Delta x)^2} \leq 1, & \frac{\sigma_2^2 \Delta t}{(\Delta y)^2} \leq 1 \\
\frac{1}{\sigma_2} \left(1\gamma - \frac{\sigma_1^2}{2} \right) - \frac{\rho\sigma_2}{\Delta x} \right) \Delta x \leq 1
\end{cases}$$

$$\frac{1}{\sigma_2^2} \left(1\gamma - \frac{\sigma_2^2}{2} \right) - \frac{\rho\sigma_1}{\Delta x} \Delta y \leq 1$$

故需要调整网格 At, Ax, Ay 来控制 d, B±. Y±, 从±20 从而. 保正 识的人作用限。意性是是作为自己。其实之 稳定性

生命至河及北京不得人生