Credit Suisse Case Study

CET-I ratio parameter estimation

```
DataB = Import["Wolfram Mathematica/Data/Credit_Suisse_Data_17-23.xlsx",
     {"Data", 2, Table[i, {i, 2, 26}], 4}];
                     表格
\texttt{DataJ} = \underbrace{\texttt{Differences}} \big[ \texttt{Tan} \big[ \texttt{Pi} * \big( \big( 1 - 2 * \texttt{DataB} \big) \, \big/ \, 2 \big) \big] \, + \, \underbrace{\texttt{Cot}} \big[ \texttt{Pi} * \big( 1 - \texttt{DataB}[[1]] \big) \big] \big] \, ;
(* solvency shocks *)
UnitRootTest[DataJ] (* stationarity check *)
单位根检验
% < 0.05
0.00073546
True
DataH = DataJ - Quantile[SmoothKernelDistribution[DataJ], 0];
                      分位数
                                 平滑核分布
(* positive transformation *)
nH = Length[DataH]
       长度
24
```

```
(* Kernel density estimation *)
kernelH = SmoothKernelDistribution[DataH];
                平滑核分布
fH[x_] := PDF[kernelH, x]; (* kernel density *)
                概率密度函数
(* Model PDFs continuous part *)
f[\lambda_{-}, \alpha_{-}, \beta_{-}, t_{-}, x_{-}] :=
   \operatorname{Exp}[(-\lambda) * t] * (\lambda * \beta^{\alpha} * t * x^{\alpha} (\alpha - 1) * (\operatorname{Exp}[(-\beta) * x] / (\alpha - 1) !) *
         {\tt HypergeometricPFQ[\{\},\ Table[1+i/\alpha,\ \{i,\ 1,\ \alpha\}],}
        广义超几何函数
                                            表格
    \lambda * t * (\beta * (x / \alpha))^{\alpha}]; (* general PDF *)
f0[\lambda_{-}, \beta_{-}, t_{-}, x_{-}] := Exp[(-\lambda) * t] * (Exp[(-\beta) * x] * Sqrt[\lambda * \beta * (t/x)] *
                                   指数形式
                                                           指数形式
         BesselI[1, 2 * Sqrt[\lambda * \beta * t * x]]); (* relaxed PDF *)
                              平方根
                                                                             概率密度函数
T = 1/4; (* quarterly frequency *)
relaxedest =
   \mathtt{Maximize}[\{\mathtt{Sum}[\mathtt{Log}[\mathtt{f0}[\lambda,\ \beta,\ \mathtt{T},\ \mathtt{DataH}[[\mathtt{i}]]]],\ \{\mathtt{i},\ \mathtt{1},\ \mathtt{nH}\}],\ \lambda>0,\ \beta>0\},
                   求和 对数
      \{\lambda, \beta\}]; (* MLE (relaxed) *)
\lambdahat0 = \lambda /. relaxedest[[2]];
\betahat0 = \beta /. relaxedest[[2]];
Print[{λhat0, βhat0}]
打印
{147.835, 37.9971}
λhat0 > 15 (* if false use a different method for reestimation *)
True
genest = For[\alpha = 1, \alpha <= 5, \alpha++,
   \{ \texttt{est}[\alpha] = \texttt{Maximize}[\{\texttt{Sum}[\texttt{Log}[\texttt{f}[\lambda, \alpha, \beta, \texttt{T}, \texttt{DataH}[[\texttt{i}]]]], \{\texttt{i}, 1, \texttt{nH}}], 
                  最大点值 求和 对数
          \lambda > 0, \beta > 0, \{\lambda, \beta\}],
   Print[est[\alpha]]}] (* MLE *)
  打印
\{1.94434, \{\lambda \rightarrow 147.835, \beta \rightarrow 37.9971\}\}
\{2.51877, \{\lambda \rightarrow 116.83, \beta \rightarrow 60.0564\}\}
\{2.79354, \{\lambda \rightarrow 106.477, \beta \rightarrow 82.1014\}\}
\{2.95453, \{\lambda \rightarrow 101.293, \beta \rightarrow 104.139\}\}
\{3.06027, \{\lambda \rightarrow 98.1804, \beta \rightarrow 126.174\}\}
```

```
\alphahat =
  Sum[If[est[\alpha][[1]] == Max[Table[est[a][[1]], {a, 1, 5}]], \alpha, 0], {\alpha, 1, 5}];
                            ... 表格
\lambda hat = Sum[If[est[\alpha][[1]] == Max[Table[est[a][[1]], \{a, 1, 5\}]],
        求和 如果
                                  表格
     \lambda /. est[\alpha][[2]], 0], {\alpha, 1, 5}];
\betahat = Sum[If[est[\alpha][[1]] == Max[Table[est[a][[1]], {a, 1, 5}]],
        求和 如果
                                  \beta /. est[\alpha][[2]], 0], {\alpha, 1, 5}];
Print[{λhat, αhat, βhat}]
打印
{98.1804, 5, 126.174}
p1 = Plot[\{fH[x], f0[\lambda hat0, \beta hat0, T, x], f[\lambda hat, \alpha hat, \beta hat, T, x]\},
    绘图
   \{x, 0, 2\}, PlotStyle \rightarrow \{Black, Dashed, Dotted\},
              绘图样式
                            黑色
                                    虚线
  PlotTheme → "Monochrome", TicksStyle → Directive["Label", 15],
                                刻度样式
                                              指令
   PlotLegends \rightarrow \{"Fitted", "Fitted (\alpha = 1)"\}, ImageSize \rightarrow Large, PlotRange \rightarrow Full] 
  绘图的图例
                                                      图像尺寸
                                                                   大
                                                                          绘制范围
3
2
1
                      0.5
                                           1.0
                                                                 1.5
                                                                                      2.0
```

Stock price parameter estimation

```
nR = Length[Rets]
   长度
```

1582

dp2 = ListLinePlot[Rets, PlotStyle -> Gray, AxesStyle -> Black, 绘制点集的线条 绘图样式 灰色 坐标轴样式

TicksStyle -> Directive["Label", 15], ImageSize -> Large, PlotRange -> Full] 指令 标签 图像尺寸 大 绘制范围 0.2 0.1 -0.1-0.2-0.3-0.4 -0.5

UnitRootTest[Rets] (* stationarity check *)

单位根检验

% < 0.05

 7.9505×10^{-19}

True

```
(* Kernel density estimation *)
kernelX = SmoothKernelDistribution[Rets];
                平滑核分布
fRets[x_] := PDF[kernelX, x]; (* kernel density *)
                     概率密度函数
(* Model PDFs (at least daily freq.) *)
fX[\lambda 1_{,} \alpha_{,} \beta_{,} \mu_{,} \sigma_{,} \lambda 2_{,} \mu V_{,} \sigma V_{,} \eta_{,} \Delta_{,} x_{]} :=
   Module [\{v = \mu + \mu V / \Delta, v = Sqrt[\sigma^2 + \sigma V^2 / \Delta]\},
   \Psi = \left(\Delta^{\wedge} \left( \left( \alpha - 1 \right) / 2 \right) / \operatorname{Sqrt} \left[ 2 * \operatorname{Pi} * \upsilon^{\wedge} 2 \right] \right) * \left( \beta * \left( \upsilon / \eta \right) \right) ^{\wedge} \alpha * \operatorname{Exp} \left[ - \left( x - \upsilon * \Delta \right) ^{\wedge} 2 / \upsilon \right]
                                    平方根   圆周率
              (2*v^2*\Delta) + (\eta*(x-v*\Delta) + \beta*v^2*\Delta)^2/((2*\eta*v)^2*\Delta)
     ParabolicCylinderD \left[-\alpha, \left(\eta * (x - v * \Delta) + \beta * v^2 * \Delta\right) / (\eta * v * Sqrt[\Delta])\right];
    抛物柱面函数
   \operatorname{Exp}\left[-\left(x-v*\Delta\right)^2/\left(2*v^2*\Delta\right)\right] + \lambda 1*\Delta*\left(1-\lambda 2*\Delta\right)*\Psi+
        指数形式
     (1 - \lambda 1 * \Delta) * ((1 - \lambda 2 * \Delta) / Sqrt[2 * Pi * \sigma^2 * \Delta]) *
         \text{Exp}\left[-\left(\mathbf{x} - \mu * \Delta\right)^2 / \left(2 * \sigma^2 * \Delta\right)\right];
        指数形式
\Delta t = 1/252; (* daily frequency *)
\lambda1hat = \lambdahat;
genretest =
 Maximize[{Sum[Log[fX[\lambda1hat, \alphahat, \betahat, \mu, \sigma, \lambda2, \muV, \sigmaV, \eta, \Deltat, Rets[[i]]]],
 最大点值
                求和 对数
       \{i, 1, nR\}], -1 < \mu < 1, 0 < \sigma < 1,
   \lambda 2 > 0, -1 < \mu V < 1, 0 < \sigma V < 1, 0 < \eta, \{\mu, \sigma, \lambda 2, \mu V, \sigma V, \eta\}
(* constrained MLE *)
\{3744.67, \{\mu \rightarrow -0.123329, \sigma \rightarrow 0.202993, 
   \lambda 2 \rightarrow 2.31387, \mu V \rightarrow 0.067659, \sigma V \rightarrow 0.0179964, \eta \rightarrow 1.79029}
\muhat = \mu /. genretest[[2]];
\sigmahat = \sigma /. genretest[[2]];
\lambda2hat = \lambda2 /. genretest[[2]];
\muVhat = \muV /. genretest[[2]];
\sigma Vhat = \sigma V /. genretest[[2]];
\etahat = \eta /. genretest[[2]];
Print[\{\mu hat, \sigma hat, \lambda 2hat, \mu Vhat, \sigma Vhat, \eta hat\}]
打印
\{-0.123329, 0.202993, 2.31387, 0.067659, 0.0179964, 1.79029\}
```

```
p2 = Plot[\{fRets[x], fX[\lambda 1hat, \alpha hat, \beta hat, \mu hat, \}]
     绘图
     \sigmahat, \lambda2hat, \muVhat, \sigmaVhat, \etahat, \Deltat, \mathbf{x}]}, {\mathbf{x}, -0.15, 0.15},
 PlotStyle -> {Black, Dashed}, PlotTheme -> "Monochrome",
                      虚线
                                绘图主题
 绘图样式
                黑色
  TicksStyle -> Directive["Label", 15],
                指令
                            标签
 PlotLegends -> {"Kernel", "Fitted"}, ImageSize -> Large, PlotRange -> Full]
                                          图像尺寸
                                                  大 绘制范围
 绘图的图例
 (* density comparison *)
                                      10
                                       5
                                                                0.10
-0.15
             -0.10
                         -0.05
                                                    0.05
                                                                             0.15
```

CoCo valuation

```
(* Distributional formulas *)
(* whole PDF of Erlang-CPP *)
             概率密度函数
fh[\lambda_{-}, \alpha_{-}, \beta_{-}, t_{-}, x_{-}] :=
    \operatorname{Exp}[(-\lambda) * t] * \left(\operatorname{DiracDelta}[x] + \lambda * \beta^{\alpha} * t * x^{\alpha} (\alpha - 1) * \left(\operatorname{Exp}[(-\beta) * x] / (\alpha - 1) !\right) * \right) 
                          狄拉克δ函数
    HypergeometricPFQ[{}, Table[1 + i / \alpha, {i, 1, \alpha}], \lambda * t * (\beta * (x / \alpha))^{\alpha}]);
   广义超几何函数
(* continuous part of PDF *)
                                 概率密度函数
fc[\lambda_{-}, \alpha_{-}, \beta_{-}, t_{-}, x_{-}] := \lambda * \beta^{\alpha} * t * x^{\alpha} (\alpha - 1) * (Exp[(-\lambda) * t - \beta * x] / (\alpha - 1)!) *
     {\tt HypergeometricPFQ[\{\},\ Table[1+i/\alpha,\ \{i,\ 1,\ \alpha\}]\,,}
                                       表格
   \lambda * t * (\beta * (x/\alpha))^\alpha;
(* time derivative of continuous part *)
tdfc[\lambda_{-}, \alpha_{-}, \beta_{-}, t_{-}, x_{-}] :=
   \lambda * \beta^{\alpha} * x^{\alpha} (\alpha - 1) * (Exp[(-\lambda) * t - \beta * x] / (\alpha - 1)!) *
                                 指数形式
  (\lambda * t * ((\beta * x) ^\alpha / Pochhammer [1 + \alpha, \alpha]) *
                             波赫汉默
          HypergeometricPFQ[\{\}, Table[2 + i/\alpha, \{i, 1, \alpha\}], \lambda * t * (\beta * (x/\alpha))^\alpha] -
                                            表格
   (\lambda * t - 1) * HypergeometricPFQ[{}, Table[1 + i/\alpha, {i, 1, \alpha}],
                    广义超几何函数
                                                      表格
            \lambda * t * (\beta * (x / \alpha))^{\alpha};
(* expectation of negative part of CCPP *)
I1[\lambda_{-}, \alpha_{-}, \beta_{-}, (t_{-})?NumericQ] := I1[\lambda, \alpha, \beta, t] =
                               数值量判定
    NIntegrate \left[ \left( 1 - \beta * (y / (\lambda * \alpha * t)) \right) * fc[\lambda, \alpha, \beta, t, y], \{y, 0, \lambda * \alpha * (t / \beta) \}, \right]
    数值积分
   AccuracyGoal -> 3];
  准确度目标
(* a rewarding probability *)
I2[\lambda_{-}, \alpha_{-}, \beta_{-}, (t_{-})] NumericQ, (x_{-}) NumericQ] := I2[\lambda_{-}, \alpha_{-}, \beta_{-}, t_{-}] =
                               数值量判定
                                                       数值量判定
     NIntegrate[tdfc[\lambda, \alpha, \beta, t, y], {y, 0, x}, AccuracyGoal -> 3];
    数值积分
                                                                        准确度目标
(* supremum probability of CCPP *)
SupPr[\lambda_, \alpha_, \beta_, t_, x_] :=
    (1 - \text{Exp}[(-\lambda) * t] - \text{NIntegrate}[fc[\lambda, \alpha, \beta, t, y], \{y, 0, x + \lambda * \alpha * (t/\beta)\}]) +
          指数形式
                               数值积分
  NIntegrate [I1[\lambda, \alpha, \beta, t - s] *
  数值积分
        ((\beta/(\lambda*\alpha))*(I2[\lambda, \alpha, \beta, s, x + \lambda*\alpha*(s/\beta)] - \lambda*Exp[(-\lambda)*s]) +
                                                                                       指数形式
    fc[\lambda, \alpha, \beta, s, x + \lambda * \alpha * (s/\beta)]), {s, 0, t}, AccuracyGoal -> 3];
                                                                          准确度目标
```

```
(* Simplified distributional formulas: \alpha=1 *)
fh0[\lambda_{-}, \beta_{-}, t_{-}, x_{-}] := Exp[(-\lambda) *t] * (DiracDelta[x] +
                                    指数形式
                                                             狄拉克δ函数
        \text{Exp}[(-\beta) * x] * \text{Sqrt}[\lambda * \beta * (t/x)] * \text{BesselI}[1, 2 * \text{Sqrt}[\lambda * \beta * t * x]]);
                              平方根
                                                             第一类修正贝塞… 平方根
\texttt{fc0}\left[\lambda_{\_},\ \beta_{\_},\ \texttt{t}_{\_},\ \texttt{x}_{\_}\right]\ :=\ \texttt{Exp}\left[\left(-\lambda\right)*\texttt{t}-\beta*\texttt{x}\right]*\texttt{Sqrt}\left[\lambda*\beta*\left(\texttt{t}/\texttt{x}\right)\right]*
                                    上指数形式
                                                                    平方根
     BesselI[1, 2 * Sqrt[\lambda * \beta * t * x]];
    第一类修正贝塞… 平方根
\mathsf{tdfc0}[\lambda_-,\ \beta_-,\ \mathsf{t}_-,\ \mathsf{x}_-]\ :=\ \lambda * \beta * \mathsf{Exp}[(-\lambda) * \mathsf{t}_- \beta * \mathsf{x}] *
                                                 指数形式
     (Hypergeometric0F1Regularized[1, \lambda * \beta * t * x] -
      正则化的合流超几何函数0F1
   \lambda * t * HypergeometricOF1Regularized[2, \lambda * \beta * t * x]);
           正则化的合流超几何函数0F1
I10[\lambda_, \beta_, (t_)?NumericQ] :=
                           数值量判定
   I1[\lambda, \beta, t] = NIntegrate \left[ \left( 1 - \beta * (x / (\lambda * t)) \right) * fc0[\lambda, \beta, t, x] \right]
                         数值积分
       \{x, 0, \lambda * (t/\beta)\}, AccuracyGoal -> 3;
                                    准确度目标
I20[\lambda_{-}, \beta_{-}, t_{-}, (x_{-})?NumericQ] := I2[\lambda, \beta, t, x] =
                                  数值量判定
     \label{eq:nonlinear} \mbox{NIntegrate[tdfc0[$\lambda$, $\beta$, t, $y$], {y, 0, x}, AccuracyGoal $->$ 3];}
                                                                        准确度目标
SupPr0[\lambda_, \beta_, t_, \mathbf{x}_] :=
    (1 - \text{Exp}[(-\lambda) * t] - \text{NIntegrate}[fc0[\lambda, \beta, t, y], \{y, 0, x + \lambda * (t/\beta)\}]) +
                                数值积分
  NIntegrate [I10[\lambda, \beta, t - s] *
  数值积分
         ((\beta/\lambda) * (I20[\lambda, \beta, s, x + \lambda * (s/\beta)] - \lambda * Exp[(-\lambda) * s]) +
            fc0[\lambda, \beta, s, x + \lambda * (s/\beta)], {s, 0, t},
   AccuracyGoal ->
   准确度目标
        3];
```

Market parameters (with artificial values)

```
W = 0.07; (* watermark *)
C0 = 0.11; (* current CET1 ratio *)
S0 = 15; (* current stock price *)
K = 10; (* principal *)
c = 6.75/2; (* semiannual coupons *)
w = 0.75; (* write-down porportion *)
r = 0.0168; (* risk-free rate *)
q = 0.02; (* dividend yield *)
T = 5; (* maturity *)
M = 2 * T; (* coupon number *)
m = 0.03; (* net profit *)
1 = 0.04; (* loan growth s.t. CET1 ratio equals watermark at infinity *)
n0 = 0.8; (* current one minus CET-1 ratio *)
(* converted trigger barrier *)
Jbar = Tan[Pi * ((1 - 2 * W) / 2)] + Cot[Pi * (1 - DataB[[1]])]
                              余切 圆周率
Jbar0 = Jbar;
2.22136
(* Govt. intervention-related functionals *)
```

```
E1[\kappa 1_{,} \varsigma 1_{,} t_{,} u_{,}] := Exp[\kappa 1^2 * (t/(2*\varsigma 1^2)) *
                                                                                                   指数形式
                         Tanh[Sqrt[2*\varsigma1^2*t^2*u]]/Sqrt[2*\varsigma1^2*t^2*u] - 1]
               Sqrt[Sech[Sqrt[2 * \varsigma1^2 * t^2]]];
             |平方根|双… |平方根
  (* continuous part expectation *)
 ac[\mu V_{-}, \sigma V_{-}, i_{-}] :=
          Which [i == 1, 1 - (1/2) * Erfc [(\mu V + \sigma V) / Sqrt[2 * \sigma V^2]], i == 2,
        Which循环
                                                                                                               补余误差函数 平方根
         (1/2) * (Erfc[(\mu V + \sigma V) / Sqrt[2 * \sigma V^2]] - Erfc[(\mu V + 2 * \sigma V) / Sqrt[2 * \sigma V^2]])
                                                                                                                                                                         补余误差函数
                                           补余误差函数
              i == 3,
         (1/2) * (Erfc[(\mu V + 2 * \sigma V) / Sqrt[2 * \sigma V^2]] -
                                                                                                             平方根
                                           补余误差函数
                           \operatorname{Erfc}[(\mu V + 3 * \sigma V) / \operatorname{Sqrt}[2 * \sigma V^2]]), i == 4,
                                                                                                平方根
                          补余误差函数
               (1/2) * Erfc[(\mu V + 3 * \sigma V) / Sqrt[2 * \sigma V^2]], True, None];
                                                                                                               平方根
                                            补余误差函数
E2[\kappa 2_{,} \varsigma 2_{,} \lambda 2_{,} \lambda 30_{,} \mu V_{,} \sigma V_{,} t_{,} u_{,}] :=
          \text{Exp}\left[(-u) * \lambda 30 * \left(\left(1 - \text{Exp}\left[\left(-\kappa 2\right) * T\right]\right) / \kappa 2\right) + \right]
                                                                                         指数形式
           \left(\lambda 2\left/\kappa 2\right)\star \text{Sum}\left[\text{ac}\left[\mu V\text{, }\sigma V\text{, i}\right]\star \text{Exp}\left[\left(-\text{i}\right)\star\varsigma 2\star\left(\text{u}\left/\kappa 2\right)\right]\star\left(\text{ExpIntegralEi}\left[\left(-\text{i}\right)\star\varsigma 2\star\left(\text{u}\left/\kappa 2\right)\right]\star\left(\text{ExpIntegralEi}\left[\left(-\text{i}\right)\star\varsigma 2\star\left(\text{u}\left/\kappa 2\right)\right]\right]\star\left(\text{ExpIntegralEi}\left[\left(-\text{i}\right)\star\varsigma 2\star\left(\text{u}\left/\kappa 2\right)\right]\right]\star\left(\text{ExpIntegralEi}\left[\left(-\text{i}\right)\star\left(\text{u}\left/\kappa 2\right)\right]\right]\star\left(\text{ExpIntegralEi}\left(-\text{i}\right)\star\left(\text{u}\left/\kappa 2\right)\right]\star\left(\text{ExpIntegralEi}\left(-\text{i}\right)\star\left(\text{u}\left/\kappa 2\right)\right)\right]\star\left(\text{ExpIntegralEi}\left(-\text{i}\right)\star\left(\text{u}\left/\kappa 2\right)\right)
                                                                                                                        指数形式
                                             i*u*(\varsigma2/\kappa2)] - ExpIntegralEi[i*u*\varsigma2*(Exp[(-\kappa2)*t]/\kappa2)]),
              \{i, 1, 4\} - \lambda 2 * t; (* jump part expectation *)
 Write-down CoCo
  (* write-off *)
 Timing[
计算时间
     WDZ = K * Exp[(-r) * T] * (1 - SupPr[\lambda hat, Ceiling[\alpha hat], \beta hat, T, Jbar]) +
                                       指数形式
                                                                                                                                                                      向上取整
        Sum[c*Exp[(-r)*k*(T/M)]*
       求和 指数形式
                         (1 - SupPr[\lambda hat, Ceiling[\alpha hat], \beta hat, k*(T/M), Jbar]), \{k, 1, M\}]
                                                                                           向上取整
 {2.96875, 41.2376}
```

```
(* write-
    down (quadrature rule applied for fast approximation with NN steps ) *)
NN = 4 * T;
\kappa1hat = 0.09;
 \kappa2hat = 3.2;
 \varsigma2hat = 0.9;
\lambda30hat = 0; (* artificial values *)
Timing WD =
计算时间
           WDZ + (1 - w) *K * Sum [Exp[(-r) * (T/NN) * j] * E1[κ1hat, ς1hat, j * (T/NN), 1] *
                                                                                     求和 指数形式
                              E2[\kappa2hat, \varsigma2hat, \lambda2hat, \lambda30hat, \muVhat, \sigmaVhat, j*(T/NN), 1]*
                (SupPr[\lambda 1hat, Ceiling[\alpha hat], \beta hat, (j + 1) * (T / NN), Jbar] -
                                                                              向上取整
                                        SupPr[\lambda1hat, Ceiling[\alphahat], \betahat, j*(T/NN), Jbar]), \{j, 1, NN\}]
                                                                                                   向上取整
 {10.2188, 41.238}
 Convertible AT1 bond pricing
  (* quadrature rule applied for fast approximation with NN steps *)
 \psi 1[p_] := (1 + \eta hat * (p / \beta hat))^{(-\alpha hat)} - 1;
 \psi^{2}[p_{-}] := \exp[\mu Vhat + \sigma Vhat^{2}/2] - 1; (* exponential correcting terms *)
Q[p_{-}] := p*q + (1-p)*r + (1/2)*p*(1-p)*\sigmahat^2 +
               \lambda 1 \text{hat} * (p * \psi 1[1] - \psi 1[p]) + \lambda 2 \text{hat} (p * \psi 2[1] - \psi 2[p]);
  (* power-adjusted dividend *)
CV[p_] :=
             WDZ + w * K * Sum [Exp[(-Q[p]) * (T/NN) * j] * E1[\kappa 1hat + p * \sigma hat * \varsigma 1hat, \varsigma 1
                                                           求和 指数形式
                                    j * (T/NN), 1] * E2[\kappa 2hat, \varsigma 2hat, \lambda 2hat * (1 + \psi 2[p]),
                 \mu Vhat + p * \sigma Vhat^2, \sigma Vhat, j * (T / NN), 1 * (SupPr[\lambda hat * (1 + \psi 1[p])),
                                             Ceiling[\alphahat], \betahat + \etahat * p, (j + 1) * (T/NN), Jbar] -
                                             向上取整
                 SupPr[\lambda hat * (1 + \psi 1[p]), Ceiling[\alpha hat], \beta hat + \eta hat * p, j * (T / NN), Jbar]),
                                                                                                                              向上取整
                          {j, 1, NN}];
```

