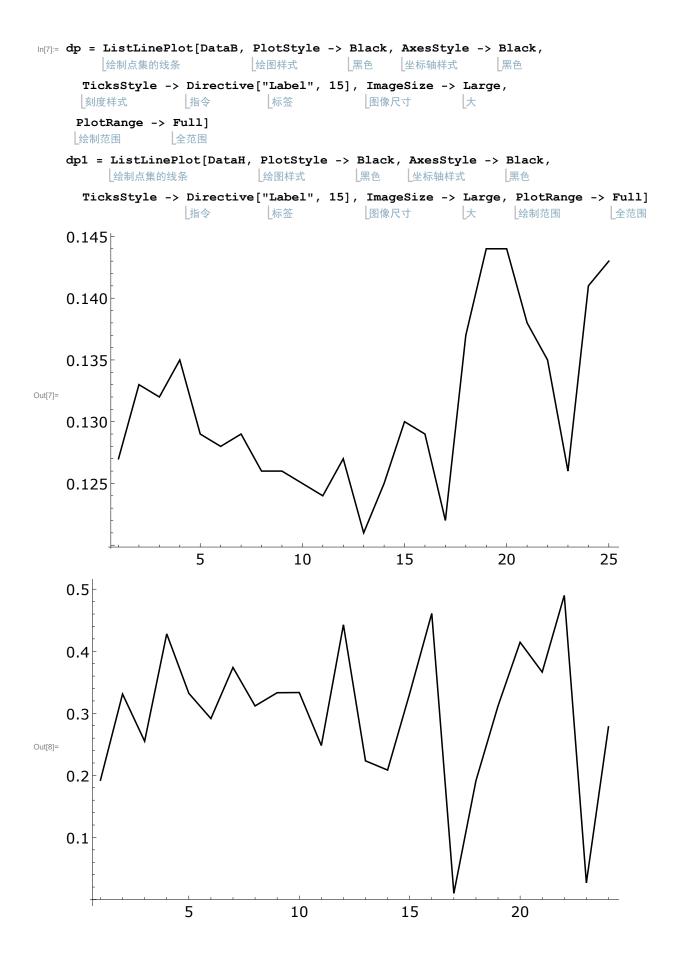
## Credit Suisse Case Study

## CET-I ratio parameter estimation



```
In[9]:= (* Kernel density estimation *)
        kernelH = SmoothKernelDistribution[DataH];
                        平滑核分布
        fH[x_] := PDF[kernelH, x]; (* kernel density *)
                        概率密度函数
In[11]:= (* 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
ln[13] := T = 1/4;
       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\},
         最大点值
                         求和 对数
           \{\lambda, \beta\}] (* MLE (relaxed) *)
Out[14]= {13.5934, {\lambda \rightarrow 32.2865, \beta \rightarrow 26.9502}}
ln[15]:= \lambda hat0 = \lambda /. relaxedest[[2]];
        \betahat0 = \beta /. relaxedest[[2]];
        Print[{λhat0, βhat0}]
       打印
        {32.2865, 26.9502}
| In[18]:= \( \lambda\) hat0 > 15 (* if false use a different method for re-estimation *)
Out[18]= True
ln[19]:= genest = For[\alpha = 1, \alpha <= 5, \alpha++,
                       For循环
            \{ \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
        \{13.5934, \{\lambda \rightarrow 32.2865, \beta \rightarrow 26.9502\}\}
        \{14.2424, \{\lambda \rightarrow 27.7388, \beta \rightarrow 46.3084\}\}
        \{14.2656, \{\lambda \rightarrow 27.1923, \beta \rightarrow 68.0939\}\}
        \{14.0225, \{\lambda \rightarrow 27.6139, \beta \rightarrow 92.1997\}\}
        \{13.6544, \{\lambda \rightarrow 28.3826, \beta \rightarrow 118.458\}\}
```

```
ln[20] = \alpha hat =
        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
                                      し… 表格
          \beta /. est[\alpha][[2]], 0], {\alpha, 1, 5}];
     Print[{λhat, αhat, βhat}]
     打印
     {27.1923, 3, 68.0939}
\ln[24] = p1 = Plot[\{fH[x], f[\lambda hat, \alpha hat, \beta hat, T, x], f0[\lambda hat0, \beta hat0, T, x]\},
           绘图
        {x, 0, 1}, PlotStyle -> {Black, Dashed, Dotted},
                    绘图样式
                                    黑色
                                             虚线
       PlotTheme -> "Monochrome", TicksStyle -> Directive["Label", 15],
       绘图主题
                                      刻度样式
                                                      指令
        PlotLegends -> {"kernel", "fitted", "fitted (\alpha = 1)"},
       绘图的图例
       ImageSize -> Large, PlotRange -> Full]
                     大
                              绘制范围
     3
     2
Out[24]=
     1
                                                                                        1.0
                      0.2
                                      0.4
                                                       0.6
                                                                        8.0
```

Stock price parameter estimation

```
In[25]:= Rets = Import["Wolfram Mathematica/Data/Credit Suisse Data 13-23.xlsx",
           导入
        {"Data", 1, Table[i, {i, 997, 2521}], 9}];
In[26]:= nR = Length[Rets]
         长度
Out[26]= 1525
In[27]:= dp2 = ListLinePlot[Rets, PlotStyle -> Gray, AxesStyle -> Black,
          绘制点集的线条
                           TicksStyle -> Directive["Label", 15], ImageSize -> Large, PlotRange -> Full]
      刻度样式 指令
                             标签
                                           图像尺寸    大    绘制范围
      0.1
Out[27]=
    -0.1
    -0.2
In[28]:= UnitRootTest[Rets] (* stationarity check *)
    单位根检验
    % < 0.05
Out[28]= 8.96902 \times 10^{-19}
Out[29]= True
     (* Kernel density estimation *)
In[30]:= kernelX = SmoothKernelDistribution[Rets];
              平滑核分布
    fRets[x_] := PDF[kernelX, x]; (* kernel density *)
                概率密度函数
```

```
In[32]:= (* 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]\},
            (2*\upsilon^2*\Delta) + (\eta*(x-v*\Delta) + \beta*\upsilon^2*\Delta)^2/((2*\eta*\upsilon)^2*\Delta)
              ParabolicCylinderD\left[-\alpha, \left(\eta * (\mathbf{x} - \vee * \Delta) + \beta * \upsilon^2 * \Delta\right) / (\eta * \upsilon * \operatorname{Sqrt}[\Delta])\right];
             抛物柱面函数
             \Psi 2 = \left(\Delta^{\left(\alpha - 1\right)/2}\right) / \operatorname{Sqrt}[2 * \operatorname{Pi} * \sigma^{2}] * (\beta * (\sigma / \eta))^{\alpha} *
                                                  平方根    圆周率
                  \operatorname{Exp}\left[-\left(\mathbf{x}-\mu\star\Delta\right)^2\left(2\star\sigma^2\star\Delta\right)+\left(\eta\star\left(\mathbf{x}-\mu\star\Delta\right)+\beta\star\sigma^2\star\Delta\right)^2\right]
                  指数形式
                 ((2*\eta*\sigma)^2*\Delta) * ParabolicCylinderD[-\alpha,
                     \left(\eta\star\left(\mathbf{x}\,-\,\mu\star\Delta\right)\,+\,\beta\star\sigma^{2}\star\Delta\right)\Big/\left(\eta\star\sigma\star\mathsf{Sqrt}[\Delta]\right)\Big]\,;
            \lambda 1 * \lambda 2 * \Delta^2 * \Psi 1 + \lambda 2 * \Delta * ((1 - \lambda 1 * \Delta) / Sqrt[2 * Pi * v^2 * \Delta]) *
                                                                              平方根    圆周率
                  \texttt{Exp} \big[ - (x - v * \Delta) ^2 / (2 * v ^2 * \Delta) \big] + \lambda 1 * \Delta * (1 - \lambda 2 * \Delta) * \Psi 2 +
                  指数形式
              (1 - \lambda 1 * \Delta) * ((1 - \lambda 2 * \Delta) / Sqrt[2 * Pi * \sigma^2 * \Delta]) *
                  \operatorname{Exp}\left[-\left(\mathbf{x}-\mu\ast\Delta\right)^2\left/\left(2\ast\sigma^2\ast\Delta\right)\right]\right];
                  指数形式
ln[33] = \Delta t = 1/252; (* daily frequency *)
In[34]:= relaxedretest =
          Maximize[\{Sum[Log[fX[\lambda hat0, 1, \beta hat0, \mu, \sigma, \lambda 2, \mu V, \sigma V, \eta, \Delta t, 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 (relaxed) *)
Out[34]= {3661.46, {\mu \rightarrow 0.186287, \sigma \rightarrow 0.225406,
             \lambda 2 \rightarrow 41.0965, \mu V \rightarrow 0.0113117, \sigma V \rightarrow 0.0343969, \eta \rightarrow 0.822461}
ln[59]:= \lambda 1hat0 = \lambda hat0;
         \muhat0 = \mu /. relaxedretest[[2]];
         \sigmahat0 = \sigma /. relaxedretest[[2]];
         \lambda 2hat0 = \lambda 2 /. relaxedretest[[2]];
         \muVhat0 = \muV /. relaxedretest[[2]];
         \sigma Vhat0 = \sigma V /. relaxedretest[[2]];
         \etahat0 = \eta /. relaxedretest[[2]];
         \texttt{Print}[\{\lambda 1 \texttt{hat0}, \ \mu \texttt{hat0}, \ \sigma \texttt{hat0}, \ \lambda 2 \texttt{hat0}, \ \mu \texttt{Vhat0}, \ \sigma \texttt{Vhat0}, \ \eta \texttt{hat0}\}]
         {32.2865, 0.186287, 0.225406, 41.0965, 0.0113117, 0.0343969, 0.822461}
```

```
ln[92]:= \lambda 1hat = \lambda hat;
       genretest =
         {\tt Maximize[\{Sum[Log[fX[\lambda 1hat, \ \alpha hat, \ \beta hat, \ \mu, \ \sigma, \ \lambda 2, \ \mu V, \ \sigma V, \ \eta, \ \Delta t, \ Rets[[i]]]],}
         最大点值 求和 对数
               \{i, 1, nR\}], -1 < \mu < 1, 0 < \sigma < 1,
          50 > \lambda 2 > 0, -1 < \mu V < 1, 0 < \sigma V < 1, 0 < \eta \}, \{\mu, \sigma, \lambda 2, \mu V, \sigma V, \eta \}]
          (* constrained MLE *)
Out[93]= {3653.83, \{\mu \to 0.382597, \sigma \to 0.235813,
           \lambda 2 \rightarrow 30.3952, \mu V \rightarrow -0.0011034, \sigma V \rightarrow 0.0539417, \eta \rightarrow 0.565409}
ln[94] = \mu hat = \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]];
        \texttt{Print}[\{\mu \texttt{hat}, \ \sigma \texttt{hat}, \ \lambda \texttt{2hat}, \ \mu \texttt{Vhat}, \ \sigma \texttt{Vhat}, \ \eta \texttt{hat}\}]
       打印
       {0.382597, 0.235813, 30.3952, -0.0011034, 0.0539417, 0.565409}
```

```
In[101]:= p2 = Plot[{fRets[x],
            绘图
          fX[\lambda 1hat, \alpha hat, \beta hat, \mu hat, \sigma hat, \lambda 2hat, \mu Vhat, \sigma Vhat, \eta hat, \Delta t, x]
         fX[\lambda 1] at 0, \frac{1}{2}, \beta hat 0, \mu hat 0, \sigma hat 0, \lambda 2 hat 0, \mu Vhat 0, \sigma Vhat 0, \eta hat 0, \Delta t, x],
         \{x, -0.15, 0.15\}, PlotStyle -> \{Black, Dashed, Dotted\},
                               绘图样式
                                                黑色 虚线
        PlotTheme -> "Monochrome", TicksStyle -> Directive["Label", 15],
        绘图主题
                                         刻度样式
                                                         指令
         PlotLegends -> {"kernel", "fitted", "fitted (\alpha = 1)"}, ImageSize -> Large,
         绘图的图例
                                                                             图像尺寸
        PlotRange -> Full] (* density comparison *)
        绘制范围
                      全范围
                                                  15
Out[101]=
                                                  10
                                                   5
      -0.15
                                                                               0.10
                                                                                             0.15
                     -0.10
                                   -0.05
                                                                 0.05
```

## CoCo valuation

```
In[102]:= (* 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 [(1 - \beta * (y / (\lambda * \alpha * t))) * fc[\lambda, \alpha, \beta, t, y], \{y, 0, \lambda * \alpha * (t / \beta)\},
            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
```

```
ln[108]:= (* Simplified distributional formulas: \alpha==1 *)
         \texttt{fh0} \left[ \lambda_{-}, \ \beta_{-}, \ \texttt{t}_{-}, \ \texttt{x}_{-} \right] \ := \ \texttt{Exp} \left[ \left( -\lambda \right) \, * \, \texttt{t} \right] \, * \, \left( \texttt{DiracDelta} \left[ \texttt{x} \right] \, + \, \right.
                                                 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]*
                                                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:nintegrate} \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)

```
In[114]:= W = 0.11; (* 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 *)
       (* converted trigger barrier *)
       Jbar = Tan[Pi * ((1 - 2 * W) / 2)] + Cot[Pi * (1 - DataB[[1]])]
                                               余切 圆周率
       Jbar0 = Jbar;
Out[114]= 0.405657
       (* Govt. intervention-related functionals *)
ln[116] = E1[\kappa 1_{,,,,} \varsigma 1_{,,,} t_{,,,} u_{,,,}] := Exp[\kappa 1^2 * (t/(2*\varsigma 1^2)) *
              \left( \text{Tanh} \left[ \text{Sqrt} \left[ 2 * \varsigma 1^2 * t^2 * u \right] \right] / \text{Sqrt} \left[ 2 * \varsigma 1^2 * t^2 * u \right] - 1 \right) \right] *
           Sqrt[Sech[Sqrt[2 * \(\zeta\)1^2 * 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,
                                                           平方根
         Which循环
                                        补余误差函数
         (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 平方根
           i == 3,
         (1/2) * (Erfc[(\mu V + 2 * \sigma V) / Sqrt[2 * \sigma V^2]] -
                    补余误差函数 平方根
               \operatorname{Erfc}[(\mu V + 3 * \sigma V) / \operatorname{Sqrt}[2 * \sigma V^2]]), i == 4,
                               平方根
           (1/2) * Erfc[(\mu V + 3 * \sigma V) / Sqrt[2 * \sigma V^2]], True, None];
                   补余误差函数 平方根
       E2[\kappa 2_{,} \varsigma 2_{,} \lambda 2_{,} \lambda 30_{,} \mu V_{,} \sigma V_{,} t_{,} u_{,}] :=
         \exp[(-u) * \lambda 30 * ((1 - \exp[(-\kappa 2) * T]) / \kappa 2) +
          (\lambda 2/\kappa 2) * Sum[ac[\mu V, \sigma V, i] * Exp[(-i) * \varsigma 2 * (u/\kappa 2)] * (ExpIntegralEi[
                    {i, 1, 4}] - \lambda 2 * t]; (* jump part expectation *)
```

```
Write-down CoCo
In[120]:= (* 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
Out[120]= \{1.42188, 34.5497\}
In[128]:= (* write-
       down (quadrature rule applied for fast approximation with NN steps ) *)
      NN = 4 * T;
      \kappa1hat = 0.09;
      \varsigma1hat = 1.4;
      \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] *
              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[\lambda 1hat, Ceiling[\alpha hat], \beta hat, j*(T/NN), Jbar]), {j, 1, NN}]
                                向上取整
Out[134]= \{5.125, 34.5725\}
```

Convertible AT1 bond pricing

```
_{\text{In}[142]:=} (* 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_] :=
                           | 求和 | 指数形式
                                              j*(T/NN), 1]*E2[\kappa2hat, \varsigma2hat, \lambda2hat*(1 + \psi^2[p]), \lambda30hat,
                             \muVhat + p * \sigmaVhat^2, \sigmaVhat, j * (T / NN), 1 | * (SupPr[\lambdahat * (1 + \psi1[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]),
                                       {j, 1, NN}];
 ln[147] = DiscretePlot[CV[p], \{p, 0, 1, 0.2\}, Joined -> True, [147] = DiscretePlot[CV[p], \{p, 0, 1, 0.2\}, Joined -> True, [147] = DiscretePlot[CV[p], [147] = DiscretePl
                                                                                                                                                  连接点
                       Filling -> None, ImageSize -> Large, PlotRange -> All]
                      填补
                                                                                  图像尺寸
                                                                                                                            大
                                                                                                                                                         绘制范围
                   34.618
                   34.617
Out[147]=
                   34.616
                   34.615
                                                                                                                                0.4
                                                                                                                                                                                 0.6
```