

FIXED INCOME SECURITIES

FRE: 6411

Sassan Alizadeh, PhD

Tandon School of Engineering

NYU

2023

Continuous Time MM

- $B(T, s_{T-1}) = B(0) \prod_{i=0}^{T-1} R(i, s_i)$
- $R(i, s_i) = (1 + r_i(s_i)\Delta t)$ where $r_i(s_i)$ is annualized Rate
- Now divide time (0,T) into "n" equal interval and $\Delta t = \frac{T}{n}$
- $B(T, s_{T-1}) = B(0) \prod_{i=0}^{n-1} \left[1 + r_i(s_i) \frac{T}{n} \right]$
- For Continuous Compounding let $\Delta t \to 0$ or $n \to \infty$
- $B(T, s_{T-1}) = B(0)e^{\int_0^T r_s ds} = e^{\int_0^T r_s ds}$
- $\frac{1}{B(T,s_{T-1})} = \frac{1}{e^{\int_0^T r_S ds}} = e^{\int_0^T -r_S ds}$
- $P(0,T) = E_0^* \left(\frac{P(T,T)=1}{B(T,S_{T-1})} \right) = E_0^* \left[e^{-\int_0^T r_S dS} \right]$

Continuous Forward Rate

- $G(t, T_1, T_2) = \frac{P(t, T_1)}{P(t, T_2)} = [1 + f(t, T_1, T_2)(\Delta T_{12})], \Delta T_{12} = T_2 T_1$
- If Continuous Compounding then :
- $f(t, T_1, T_2) = \frac{1}{\Delta T_{12}} \left[\ln P(t, T_1) \ln P(t, T_2) \right]$
- $f(t,T) = \lim_{\Delta T_{12} \to 0} \frac{1}{\Delta T_{12}} \left[\ln P(t,T_1) \ln P(t,T_2) \right] = -\frac{\partial \ln P(t,T)}{\partial T}$
- $p(t,T) = e^{-\int_0^T f(t,s)ds}$
- $p(t,T) = e^{-\int_0^T f(t,s)ds} = E_0^* \left[e^{-\int_0^T r_s ds} \right]$
- Where $f(t,t) = r_t$



$$P(0,T) = E_0^* \left(\frac{P(T,T)=1}{B(T,S_{T-1})} \right) = E_0^* \left[\exp(-\int_0^T r_t dt) \right]$$

•
$$E(0,T) = E_0(\frac{P(T,T)=1}{B(T,S_{T-1})}) = E_0\left[\exp(-\int_0^T r_t dt)\right]$$

- Here E(0,T) is the time 0 expected value of \$1 at T
- Here P(0,T) is the time 0 value of \$1 at T
- Note for P(0,T) we use the pseudo-probability or the risk
 adjusted probability where as for E(0,T) we use the real probability

Local Expectation Hypothesis

From the above equations we get YTM using price P(0,T):

$$Y(0,T) = P(0,T)^{\frac{-1}{T}} = E_0^* \left(\frac{1}{B(T,S_{T-1})}\right)^{\frac{-1}{T}}$$

$$Y(0,T) = -\frac{\ln(P(0,T))}{T} = -\frac{\ln(E_0^* \left[exp(-\int_0^T r_t dt)\right]}{T}$$

Consider find eYTM using E(0,T) instead of P(0,T)

•
$$eY(0,T) = E(0,T)^{\frac{-1}{T}} = E_0 \left(\frac{1}{B(T,S_{T-1})}\right)^{\frac{-1}{T}}$$

•
$$eY(0,T) = -\frac{\ln(E(0,T))}{T} = -\frac{\ln(E_0[exp(-\int_0^T r_t dt)]}{T}$$

Local Expectation Hypothesis

•
$$P(0,T)=E_0^*\left(\frac{P(T,T)}{B(T,S_{T-1})}\right)=E_0^*\left[\exp(-\int_0^T r_t dt)\right]$$

•
$$E(0,T) = E_0(\frac{P(T,T)}{B(T,S_{T-1})}) = E_0\left[\exp(-\int_0^T r_t dt)\right]$$

$$Y(0,T) = -\frac{\ln(P(0,T))}{T} = -\frac{\ln(E_0^* \left[exp(-\int_0^T r_t dt) \right]}{T} = \frac{1}{T} \int_0^T f(t,s) \, ds$$

•
$$eY(0,T) = -\frac{\ln(E(0,T))}{T} = -\frac{\ln(E_0[exp(-\int_0^T r_t dt)]}{T}$$



Local Expectation Hypothesis

• LEH:
$$y(0,T) = ey(0,T)$$

- That is the expectation under the pseudo-probability and actual probability is the same.
- Actual probability is the same as risk-adjusted probabilities. (risk neutrality)
- We can show under LEH:

$$E_t\left(\frac{P(t+1,T,S_{t+1})}{P(t,T,S_t)}\right) = E_t^*\left(\frac{P(t+1,T,S_{t+1})}{P(t,T,S_t)}\right) = r(t,S_t)$$

 The expected actual return is the same as expected risk adjusted return which is risk free rate.

Expectation Hypothesis

Expectation Hypothesis:

$$EH: y(0,T) = \frac{E((R_0-1)+\cdots+(R(T-1,S_{T-2})-1))}{T} = \frac{1}{T}E\left(\int_0^T r_t dt\right)$$

$$Y(0,T) = -\frac{\ln(P(0,T))}{T} = \frac{1}{T} \int_0^T f(t,s) \, ds = \frac{1}{T} E\left(\int_0^T r_s \, ds\right)$$
$$\int_0^T E[f(t,s) - r_s] \, ds = 0$$

Note the difference between the two hypothesis

Or YTM is the expected average of the short rates

Expectation Hypothesis

• LEH:
$$y(0,T) = -\frac{\ln(E_0[exp(-\int_0^T r_t dt)]}{T}$$

$$EH: y(0,T) = \frac{1}{T}E_0\left(\int_0^T r_t dt\right)$$

By Jensen's Ineqaulity we know:

$$-\frac{\ln(E\left[exp(-\int_0^T r_t dt)\right]}{T} <= -\frac{E(\ln\left[exp(-\int_0^T r_t dt)\right]}{T} = \frac{E(\int_0^T r_t dt)}{T}$$

So if LEH hold then EH usually will not hold