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LAMPIRAN

Lampiran 1 : Pembuktian Persamaan 2.39 dan 2.40

Pembuktian dari Triana,S (2017) sebagai berikut
dari persamaan 2.34 dan 2.38 diperoleh persamaan sebagai berikut.

$$\begin{aligned} E(C_{i,n-i}|C_{i,n}) &= p_i C_{i,n} \\ Var(C_{i,n-i}|C_{i,n}) &= p_i q_i \alpha_i^2 (C_{i,n}) \end{aligned}$$

Dengan demikian, akan diperoleh bentuk $Var(C_{i,n-i})$ dan $Cov(C_{i,n-i}, C_{i,n})$ sebagai berikut.

$$\begin{aligned} Var(C_{i,n-i}) &= E[Var[C_{i,n-i}|C_{i,n}]] + Var[E[C_{i,n-i}|C_{i,n}]] \\ &= p_i q_i E[\alpha_i^2(C_{i,n})] + p_i^2 Var[C_{i,n}] \\ &= p_i(1-p_i)E[\alpha_i^2(C_{i,n})] + p_i^2 Var[C_{i,n}] \\ &= p_i E[\alpha_i^2(C_{i,n})] - p_i^2 E[\alpha_i^2(C_{i,n})] + p_i^2 Var[C_{i,n}] \\ Var(C_{i,n-i}) &= p_i E[\alpha_i^2(C_{i,n})] + p_i^2 (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \end{aligned}$$

$$\begin{aligned} Cov(C_{i,n-i}, C_{i,n}) &= E[Cov[C_{i,n-i}, C_{i,n}|C_{i,n}]] + Cov[E[C_{i,n-i}|C_{i,n}], E[C_{i,n}|C_{i,n}]] \\ &= E[Cov[C_{i,n-i}, C_{i,n}|C_{i,n}]] + Cov[E[C_{i,n-i}|C_{i,n}], C_{i,n}] \\ &= E[E(C_{i,n-i}C_{i,n}|C_{i,n}) - E(C_{i,n-i}|C_{i,n})E(C_{i,n}|C_{i,n})] + Cov[p_i C_{i,n}, C_{i,n}] \\ &= E[E(C_{i,n-i}C_{i,n}|C_{i,n}) - p_i C_{i,n} C_{i,n}] + p_i Cov[C_{i,n}, C_{i,n}] \\ &= E[E(C_{i,n-i}C_{i,n}|C_{i,n})] - p_i E[C_{i,n}^2] + p_i Var[C_{i,n}] \\ &= E[C_{i,n-i}C_{i,n}] - p_i E[C_{i,n}^2] + p_i Var[C_{i,n}] \\ &= E[p_i C_{i,n} C_{i,n}] - p_i E[C_{i,n}^2] + p_i Var[C_{i,n}] \\ &= p_i E[C_{i,n}^2] - p_i E[C_{i,n}^2] + p_i Var[C_{i,n}] \\ Cov(C_{i,n-i}, C_{i,n}) &= p_i Var[C_{i,n}] \end{aligned}$$

Subtitusikan $Var(C_{i,n-i})$ dan $Cov(C_{i,n-i}, C_{i,n})$ maka didapatkan nilai $Cov(C_{i,n-i}, R_i)$ sebagai berikut

$$\begin{aligned}
 Cov[C_{i,n-i}, R_i] &= Cov[C_{i,n-i}, C_{i,n} - C_{i,n-i}] = Cov[C_{i,n-i}, C_{i,n}] - Cov[C_{i,n-i}, C_{i,n-i}] \\
 &= Cov[C_{i,n-i}, C_{i,n}] - Var[C_{i,n-i}] \\
 &= p_i Var[C_{i,n}] - p_i E[\alpha_i^2(C_{i,n})] - p_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \\
 &= p_i Var[C_{i,n}] - p_i E[\alpha_i^2(C_{i,n})] - p_i(1 - q_i)(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \\
 &= p_i Var[C_{i,n}] - p_i E[\alpha_i^2(C_{i,n})] - (p_i - p_i q_i)(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \\
 &= p_i Var[C_{i,n}] - p_i E[\alpha_i^2(C_{i,n})] - p_i Var[C_{i,n}] + p_i q_i Var[C_{i,n}] \\
 &\quad + p_i E[\alpha_i^2(C_{i,n})] - p_i q_i E[\alpha_i^2(C_{i,n})]
 \end{aligned}$$

$$Cov[C_{i,n-i}, R_i] = p_i q_i (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})])$$

Maka dengan mensubtitusikan $Var(C_{i,n-i})$ dan $Cov[C_{i,n-i}, R_i]$ ke persamaan didapatkan

$$\begin{aligned}
 c_i^* &= \frac{p_i}{q_i} \frac{Cov[C_{i,n-i}, R_i] + p_i q_i Var[U_i^{BC}]}{Var[C_{i,n-i}] + p_i^2 Var[U_i^{BC}]} \\
 &= \frac{p_i}{q_i} \frac{p_i q_i (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) + p_i q_i Var[U_i^{BC}]}{p_i E[\alpha_i^2(C_{i,n})] + p_i^2 (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) + p_i^2 Var[U_i^{BC}]} \\
 &= p_i \frac{Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})] + Var[U_i^{BC}]}{E[\alpha_i^2(C_{i,n})] + p_i (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) + p_i Var[U_i^{BC}]} \\
 &= \frac{p_i (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})] + Var[U_i^{BC}])}{p_i (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})] + Var[U_i^{BC}]) + E[\alpha_i^2(C_{i,n})]} \\
 &= \frac{p_i}{\frac{E[\alpha_i^2(C_{i,n})]}{p_i + \frac{Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})] + Var[U_i^{BC}]}{Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})] + Var[U_i^{BC}]}}}
 \end{aligned}$$

$$c_i^* = \frac{p_i}{p_i + t_i^*}$$

Dengan $t_i^* = \frac{E[\alpha_i^2(C_{i,n})]}{Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})] + Var[U_i^{BC}]}$ maka persamaan 3.39 dan 3.40 Terbukti

Lampiran 2 : Pembuktian Persamaan 3.44

Pembuktian dari Triana,S (2017) sebagai berikut:

Subtitusikan persamaan 3.41 Serta asumsi $Var[C_{i,n}] = Var[U_i^{BC}]$ dan $E[C_{i,n}] = E[U_i^{BC}]$ pada $Var(C_{i,n-i})$ yang telah didapat sebelumnya. Sehingga akan diperoleh

bentuk $Var(C_{i,n-i})$ yang lain

$$\begin{aligned} Var(C_{i,n-i}) &= p_i E[\alpha_i^2(C_{i,n})] + p_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \\ &= p_i \beta_i^2 (Var[U_i^{BC}] + (E[U_i^{BC}])^2) \\ &\quad + p_i^2 (Var[U_i^{BC}] - \beta_i^2 (Var[U_i^{BC}] + (E[U_i^{BC}])^2)) \\ &= p_i \beta_i^2 Var[U_i^{BC}] + p_i \beta_i^2 (E[U_i^{BC}])^2 + p_i^2 Var[U_i^{BC}] - p_i^2 \beta_i^2 Var[U_i^{BC}] \\ &\quad - p_i^2 \beta_i^2 (E[U_i^{BC}])^2 \\ &= Var[U_i^{BC}] \left(p_i \beta_i^2 + p_i^2 - p_i^2 \beta_i^2 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} [p_i \beta_i^2 - p_i^2 \beta_i^2] \right) \\ Var(C_{i,n-i}) &= Var[U_i^{BC}] \left(p_i [\beta_i^2(1-p_i) + p_i] + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} [p_i \beta_i^2(1-p_i)] \right) \end{aligned}$$

Dengan menggunakan asumsi $Var[C_{i,n}] = Var[U_i^{BC}]$ dan $E[C_{i,n}] = E[U_i^{BC}]$ dan persamaan 3.41 pada persamaan 3.40 maka akan diperoleh estimasi dari t_i sebagai berikut

$$\begin{aligned} t_i &= \frac{E[\alpha_i^2(C_{i,n})]}{Var[U_i^{BC}] + Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]} \\ &= \frac{\beta_i^2 (Var[U_i^{BC}] + (E[U_i^{BC}])^2)}{Var[U_i^{BC}] + Var[U_i^{BC}] - \beta_i^2 (Var[U_i^{BC}] + (E[U_i^{BC}])^2)} \\ &= \frac{\beta_i^2 (Var[U_i^{BC}] + (E[U_i^{BC}])^2)}{2Var[U_i^{BC}] - \beta_i^2 (Var[U_i^{BC}] + (E[U_i^{BC}])^2)} \frac{\frac{1}{Var[U_i^{BC}]}}{\frac{1}{Var[U_i^{BC}]}} \\ t_i &= \frac{\beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right)}{2 - \beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right)} \end{aligned}$$

sehingga diperoleh

$$t_i \left(2 - \beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right) \right) = \beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right)$$

$$2t_i - t_i \beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right) = \beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right)$$

$$(1 + t_i) \beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right) = 2t_i$$

$$\beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right) = \frac{2t_i}{(1 + t_i)}$$

Maka dengan mensubtitusikan nilai $Var(C_{i,n-i})$ dan $\beta_i^2 \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right)$ yang telah

didapat ke dalam persamaan 3.43 maka didapatkan

$$\begin{aligned} Var[R_i^B] &= c_i^2 \left(\frac{q_i}{p_i} \right)^2 Var[C_{i,n-i}] + (1 - c_i)^2 q_i^2 Var[U_i^{BC}] \\ &= c_i^2 \left(\frac{q_i}{p_i} \right)^2 Var[U_i^{BC}] \left(p_i [\beta_i^2 (1 - p_i) + p_i] + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} [p_i \beta_i^2 (1 - p_i)] \right) \\ &\quad + (1 - c_i)^2 Var[R_i^{BF}] \\ &= Var[R_i^{BF}] \left(c_i^2 \left(\frac{1}{p_i} [\beta_i^2 (1 - p_i) + p_i] + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \left[\frac{1}{p_i} \beta_i^2 (1 - p_i) \right] \right) \right. \\ &\quad \left. + (1 - c_i)^2 \right) \\ &= Var[R_i^{BF}] \left(c_i^2 \left(\beta_i^2 \frac{(1-p_i)}{p} + 1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \left[\beta_i^2 \frac{(1-p_i)}{p_i} \right] \right) + (1 - c_i)^2 \right) \\ &= Var[R_i^{BF}] \left(c_i^2 \left(1 + \beta_i^2 \frac{(1-p_i)}{p} \left(1 + \frac{(E[U_i^{BC}])^2}{Var[U_i^{BC}]} \right) \right) + (1 - c_i)^2 \right) \\ &= Var[R_i^{BF}] \left(c_i^2 \left(1 + \frac{(1-p_i)}{p_i} \left(\frac{2t_i}{(1+t_i)} \right) \right) + c_i^2 - 2c_i + 1 \right) \\ &= Var[R_i^{BF}] \left(c_i^2 + 2c_i^2 \left(\frac{1-p_i}{p_i} \left(\frac{t_i}{1+t_i} \right) \right) + c_i^2 - 2c_i + 1 \right) \\ &= Var[R_i^{BF}] \left(2c_i^2 + 2c_i^2 \left(\frac{1-p_i}{p_i} \left(\frac{t_i}{1+t_i} \right) \right) - 2c_i + 1 \right) \\ &= Var[R_i^{BF}] \left(2c_i^2 \left[1 + \left(\frac{1-p_i}{p_i} \left(\frac{t_i}{1+t_i} \right) \right) - 2c_i + 1 \right] \right) \end{aligned}$$

$$\begin{aligned}
 &= Var[R_i^{BF}] \left(2 \left(\frac{p_i}{p_i+t_i} \right)^2 \left[1 + \left(\frac{1-p_i}{p_i} \right) \left(\frac{t_i}{1+t_i} \right) \right] - 2 \frac{p_i}{p_i+t_i} + 1 \right) \\
 &= Var[R_i^{BF}] \left(\frac{2p_i^2}{(p_i+t_i)^2} + \frac{2p_i(1-p_i)}{(p_i+t_i)^2} \left(\frac{t_i}{1+t_i} \right) - \frac{2p_i}{p_i+t_i} + 1 \right) \\
 &= Var[R_i^{BF}] \left(\frac{2p_i^2(1+t_i) + 2p_i t_i (1-p_i) - 2p_i(p_i+t_i)(1+t_i)}{(p_i+t_i)^2(1+t_i)} + 1 \right) \\
 &= Var[R_i^{BF}] \left(\frac{2p_i^2 + 2p_i^2 t_i + 2p_i t_i - 2p_i^2 t_i - 2p_i^2 - 2p_i^2 t_i - 2p_i t_i - 2p_i t_i^2}{(p_i+t_i)^2(1+t_i)} + 1 \right) \\
 &= Var[R_i^{BF}] \left(\frac{-2p_i^2 t_i - 2p_i t_i^2}{(p_i+t_i)^2(1+t_i)} + 1 \right) \\
 &\quad Var[R_i^{BF}] \left(1 - \frac{(2p_i t_i (p_i+t_i))}{(p_i+t_i)^2(1+t_i)} \right) \\
 Var[R_i^B] &= Var[R_i^{BF}] \left(1 - \frac{2p_i t_i}{(p_i+t_i)(1+t_i)} \right)
 \end{aligned}$$

Apabila $Var[R_i^B]$ diatas diturunkan terhadap t_i , maka diperoleh t_i yang meminimumkan variansi sebagai berikut

$$\begin{aligned}
 \frac{\partial}{\partial t_i} Var[R_i^B] &= \frac{\partial}{\partial t_i} \left[Var[R_i^{BF}] \left(1 - \frac{2p_i t_i}{(p_i+t_i)(1+t_i)} \right) \right] \\
 0 &= \frac{\partial}{\partial t_i} \left[Var[R_i^{BF}] \left(1 - \frac{2p_i t_i}{(p_i+t_i)(1+t_i)} \right) \right] \\
 0 &= \frac{\partial}{\partial t_i} \left[\frac{2p_i t_i}{(p_i + p_i t_i + t_i + t_i^2)} \right] \\
 0 &= \frac{2p_i(p_i + p_i t_i^* + t_i^* + t_i^{*2}) - (p_i + 1 + 2t_i^*)(2p_i t_i^*)}{(p_i + p_i t_i^* + t_i^* + t_i^{*2})^2} \\
 0 &= 2p_i^2 + 2p_i^2 t_i^* + 2p_i t_i^* + 2p_i t_i^{*2} - 2p_i^2 t_i^* - 2p_i t_i^* - 4p_i t_i^{*2} \\
 0 &= 2p_i^2 - 2p_i t_i^{*2} \\
 t_i^{*2} &= \frac{2p_i^2}{2p_i} \\
 t_i^* &= \sqrt{p_i}
 \end{aligned}$$

Karena $p_i \neq 0$ untuk semua i dengan demikian t_i yang meminimumkan variansi $Var[R_i^B]$ adalah $t_i^* = \sqrt{p_i}$. dengan demikian persamaan 3.44 terbukti.

Lampiran 3 : pembuktian persamaan 3.46

Pembuktian dari Triana,S (2017) sebagai berikut :

Sebelum membuktikan persamaan 3.46 terlebih dulu akan dicari nilai $Var[R_i]$ menggunakan persamaan $Var(C_{i,n-i})$ dan $Cov(C_{i,n-i}, C_{i,n})$ yang telah didapat pada lampiran 1 sebagai berikut

$$Var(C_{i,n-i}) = p_i E[\alpha_i^2(C_{i,n})] + p_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})])$$

$$Cov(C_{i,n-i}, C_{i,n}) = p_i Var[C_{i,n}]$$

Maka nilai $Var[R_i]$ adalah

$$\begin{aligned} Var[R_i] &= Var[C_{i,n} - C_{i,n-i}] \\ &= Var[C_{i,n}] - 2Cov[C_{i,n} - C_{i,n-i}] + Var[C_{i,n-i}] \\ &= Var[C_{i,n}] - 2(p_i Var[C_{i,n-i}]) + p_i E[\alpha_i^2(C_{i,n})] + p_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \\ &= Var[C_{i,n}] - 2p_i Var[C_{i,n}] + p_i E[\alpha_i^2(C_{i,n})] + p_i^2 Var[C_{i,n}] - p_i^2 E[\alpha_i^2(C_{i,n})] \\ &= Var[C_{i,n}](1 - 2p_i + p_i^2) + p_i E[\alpha_i^2(C_{i,n})](1 - p_i) \\ &= Var[C_{i,n}](1 - p_i)^2 + p_i q_i E[\alpha_i^2(C_{i,n})] \\ &= q_i^2 Var[C_{i,n}] + q_i(1 - q_i)E[\alpha_i^2(C_{i,n})] \\ &= q_i^2 Var[C_{i,n}] + (q_i - q_i^2)E[\alpha_i^2(C_{i,n})] \\ Var[R_i] &= q_i E[\alpha_i^2(C_{i,n})] + q_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \end{aligned}$$

a. MSE Chain Ladder

Selanjutnya dengan asumsi $E[R_{CL}] = E[R_i]$ diperoleh mean square error metode *Chain Ladder* sebagai berikut

$$\begin{aligned} mse(R_i^{CL}) &= E[(R_i^{CL} - R_i)^2] \\ &= Var[R_i^{CL} - R_i] \\ &= Var[R_i^{CL}] - 2 Cov[R_i^{CL}, R_i] + Var[R_i] \\ &= \left(\frac{q_i}{p_i}\right)^2 Var[C_{i,n-i}] - 2 \frac{q_i}{p_i} Cov[C_{i,n-i}, R_i] + Var[R_i] \\ &= \left(\frac{q_i}{p_i}\right)^2 (p_i E[\alpha_i^2(C_{i,n})] + p_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})])) - \\ &\quad 2 \frac{q_i}{p_i} (p_i, q_i(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})])) + q_i E[\alpha_i^2(C_{i,n})] + \\ &\quad q_i^2(Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \end{aligned}$$

$$\begin{aligned}
&= \frac{q_i^2}{p_i} E[\alpha_i^2(C_{i,n})] + q_i^2 Var[C_{i,n}] - q_i^2 E[\alpha_i^2(C_{i,n})] - \\
&\quad 2q_i^2 Var[C_{i,n}] + 2q_i^2 E[\alpha_i^2(C_{i,n})] + q_i E[\alpha_i^2(C_{i,n})] + \\
&\quad q_i^2 Var[C_{i,n}] - q_i^2 E[\alpha_i^2(C_{i,n})] \\
&= E[\alpha_i^2(C_{i,n})] \left(\frac{q_i^2}{p_i} + q_i \right) \\
&= E[\alpha_i^2(C_{i,n})] \left(\frac{q_i^2 + q_i p_i}{p_i} \right) \\
&= E[\alpha_i^2(C_{i,n})] \left(\frac{q_i^2 + q_i(1-p_i)}{p_i} \right) \\
mse(R_i^{CL}) &= E[\alpha_i^2(C_{i,n})] \frac{q_i}{p_i}
\end{aligned}$$

b. MSE Bornhuetter Ferguson

Dari asumsi sebelumnya, diperoleh bahwa $E[R_i^{BF}] = q_i E[U_i^{BC}] = q_i E[C_{i,n}] = E[C_{i,n} - P_i C_{i,n}^{CL}] = E[C_{i,n} - C_{i,n-i}] = E[R_i]$. Asumsikan bahwa $Cov[R_i^{BF}, R_i] = q_i Cov[U_i^{BC}, R_i] = 0$. Dengan demikian, diperoleh *mean squared error* dari metode Bornhuetter Ferguson sebagai berikut

$$\begin{aligned}
mse(R_i^{BF}) &= E[(R_i^{BF} - R_i)^2] \\
&= Var[R_i^{BF} - R_i] \\
&= Var[R_i^{BF}] + Var[R_i] \\
&= q_i^2 Var[U_i^{BC}] + q_i^2 (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) + q_i E[\alpha_i^2(C_{i,n})] \\
&= q_i E[\alpha_i^2(C_{i,n})] + q_i^2 (Var[U_i^{BC}] + Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \\
&= E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i (Var[U_i^{BC}] + Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]))}{E[\alpha_i^2(C_{i,n})]} \right) \\
&= E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{\frac{q_i}{E[\alpha_i^2(C_{i,n})]} }{\frac{Var[U_i^{BC}] + Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]}{E[\alpha_i^2(C_{i,n})]}} \right) \\
mse(R_i^{BF}) &= E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i}{t_i} \right)
\end{aligned}$$

c. MSE Benktander

$$\begin{aligned}
mse(R_i^{GB}) &= E \left[\left(c_i(R_i^{CL} - R_i) + (1 - c_i)(R_i^{BF} - R_i) \right)^2 \right] \\
&= c_i^2 mse(R_i^{CL}) + 2c_i(1 - c_i)E[(R_i^{CL} - R_i)(R_i^{BF} - R_i)] \\
&\quad + (1 - c_i)^2 mse(R_i^{BF}) \\
&= c_i^2 E \left[\alpha_i^2(C_{i,n}) \frac{q_i}{p_i} \right] + 2c_i(1 - c_i)Cov[R_i^{CL} - R_i, R_i^{BF} - R_i] \\
&\quad + (1 - c_i)^2 E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i}{t_i} \right) \\
&= c_i^2 E[\alpha_i^2(C_{i,n})] \frac{q_i}{p_i} + 2c_i(1 - c_i)(Var[R_i] - Cov[R_i^{CL}, R_i]) + \\
&\quad (1 - c_i)^2 E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i}{t_i} \right) \\
&= c_i^2 E[\alpha_i^2(C_{i,n})] \frac{q_i}{p_i} + 2c_i(1 - c_i) \left(q_i E[\alpha_i^2(C_{i,n})] + q_i^2 (Var[C_{i,n}] - \right. \\
&\quad \left. E[\alpha_i^2(C_{i,n})]) - \frac{q_i}{p_i} Cov[C_{i,n-i}, R_i] \right) + (1 - c_i)^2 E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i}{t_i} \right) \\
&= c_i^2 E[\alpha_i^2(C_{i,n})] \frac{q_i}{p_i} + 2c_i(1 - c_i) \left(q_i E[\alpha_i^2(C_{i,n})] + q_i^2 (Var[C_{i,n}] - \right. \\
&\quad \left. E[\alpha_i^2(C_{i,n})]) - \frac{q_i}{p_i} q_i p_i (Var[C_{i,n}] - E[\alpha_i^2(C_{i,n})]) \right) + (1 - c_i)^2 E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i}{t_i} \right) \\
&= c_i^2 E[\alpha_i^2(C_{i,n})] + 2c_i(1 - c_i)(q_i E[\alpha_i^2(C_{i,n})]) + (1 - c_i)^2 E[\alpha_i^2(C_{i,n})] q_i \left(1 + \frac{q_i}{t_i} \right) \\
&= E[\alpha_i^2(C_{i,n})] \left[c_i^2 \frac{q_i}{p_i} + 2c_i(1 - c_i)(q_i) + (1 - c_i)^2 q_i \left(1 + \frac{q_i}{t_i} \right) \right] \\
&= E[\alpha_i^2(C_{i,n})] \left[c_i^2 \frac{q_i}{p_i} + 2c_i q_i - 2c_i^2 q_i + (1 - c_i)^2 q_i + (1 - c_i)^2 \frac{q_i^2}{t_i} \right] \\
&= E[\alpha_i^2(C_{i,n})] \left[c_i^2 \frac{q_i}{p_i} + 2c_i q_i - 2c_i^2 q_i + q_i - 2c_i q_i + c_i^2 q_i + \right. \\
&\quad \left. (1 - c_i)^2 \frac{q_i^2}{t_i} \right]
\end{aligned}$$

$$\begin{aligned}
 &= E[\alpha_i^2(C_{i,n})] \left[c_i^2 \left(\frac{q_i}{p_i} - q_i \right) + q_i + (1 - c_i)^2 \frac{q_i^2}{t_i} \right] \\
 &= E[\alpha_i^2(C_{i,n})] \left[c_i^2 \left(\frac{q_i - q_i(1-q_i)}{p_i} \right) + q_i + (1 - c_i)^2 \frac{q_i^2}{t_i} \right] \\
 &= E[\alpha_i^2(C_{i,n})] \left[c_i^2 \left(\frac{q_i^2}{p_i} \right) + q_i + (1 - c_i)^2 \frac{q_i^2}{t_i} \right] \\
 mse(R_i^{GB}) &= E[\alpha_i^2(C_{i,n})] \left(\frac{c_i^2}{p_i} + \frac{1}{q_i} + \frac{(1-c_i)^2}{t_i} \right) q_i^2
 \end{aligned}$$

Dengan demikian persamaan 3.46 terbukti.