

# Term Structure Models

Caleb Vinson

## Models

Table 1: Comparison of Interest Rate Models

Model	Uniqueness	Pro	Con
<b>Vasicek</b>	Mean-reverting with constant volatility.	Analytical tractability.	Allows negative rates.
<b>CIR</b>	Non-negativity via square-root volatility.	Ensures positive rates.	Difficult to calibrate.
<b>Hull-White</b>	Time-dependent mean reversion.	Fits initial yield curve.	Increased calibration complexity.
<b>Black-Karasinski</b>	Logarithmic modeling of the short rate.	Ensures positive rates.	Requires numerical solutions.
<b>Nelson-Siegel</b>	Empirical yield curve representation.	Intuitive interpretation of parameters.	Lacks arbitrage-free properties.

## Vasicek Model

### Derivation of the Vasicek Model

The Vasicek model is a one-factor short-rate model described by the following stochastic differential equation (SDE):

$$dr_t = \kappa(\theta - r_t)dt + \sigma dW_t,$$

where: -  $r_t$  is the short rate at time  $t$ , -  $\kappa > 0$  is the speed of mean reversion, -  $\theta$  is the long-term mean level of the short rate, -  $\sigma > 0$  is the volatility of the short rate, -  $W_t$  is a standard Brownian motion.

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### 1. Mean-Reverting Property

The term  $\kappa(\theta - r_t)$  represents the mean-reverting drift, ensuring that the short rate tends to revert to the long-term mean  $\theta$  over time at a speed determined by  $\kappa$ .

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## 2. Solution of the SDE

To solve the SDE, we use the method of integrating factors. Let us rewrite the SDE:

$$dr_t + \kappa r_t dt = \kappa \theta dt + \sigma dW_t.$$

Multiply through by  $e^{\kappa t}$ , the integrating factor:

$$e^{\kappa t} dr_t + \kappa e^{\kappa t} r_t dt = \kappa \theta e^{\kappa t} dt + \sigma e^{\kappa t} dW_t.$$

The left-hand side becomes an exact differential:

$$d(e^{\kappa t} r_t) = \kappa \theta e^{\kappa t} dt + \sigma e^{\kappa t} dW_t.$$

Integrating both sides from 0 to  $t$ :

$$e^{\kappa t} r_t - r_0 = \int_0^t \kappa \theta e^{\kappa s} ds + \int_0^t \sigma e^{\kappa s} dW_s.$$

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## 3. Simplifying the Deterministic Term

The deterministic integral is:

$$\int_0^t \kappa \theta e^{\kappa s} ds = \theta (e^{\kappa t} - 1).$$

---

## 4. Simplifying the Stochastic Term

The stochastic integral remains as:

$$\int_0^t \sigma e^{\kappa s} dW_s.$$

For now, we leave it as is.

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## 5. Final Solution

Substituting back, we get:

$$r_t = r_0 e^{-\kappa t} + \theta (1 - e^{-\kappa t}) + \sigma \int_0^t e^{-\kappa(t-s)} dW_s.$$

This is the explicit solution to the Vasicek model. The first term represents the decay of the initial short rate, the second term represents the drift toward the mean level  $\theta$ , and the third term represents the stochastic component driven by Brownian motion.

```
# Vasicek

# Parameters

# Initial int rate (r0)
r0 <- 0.03

# Mean Reversion
a <- 0.1
b <- 0.05

# volatility
sigma <- 0.02

T <- 1

dt <- 1/252
n <- T/dt

# Starting Point
short_rate <- numeric(n)
short_rate[1] <- r0

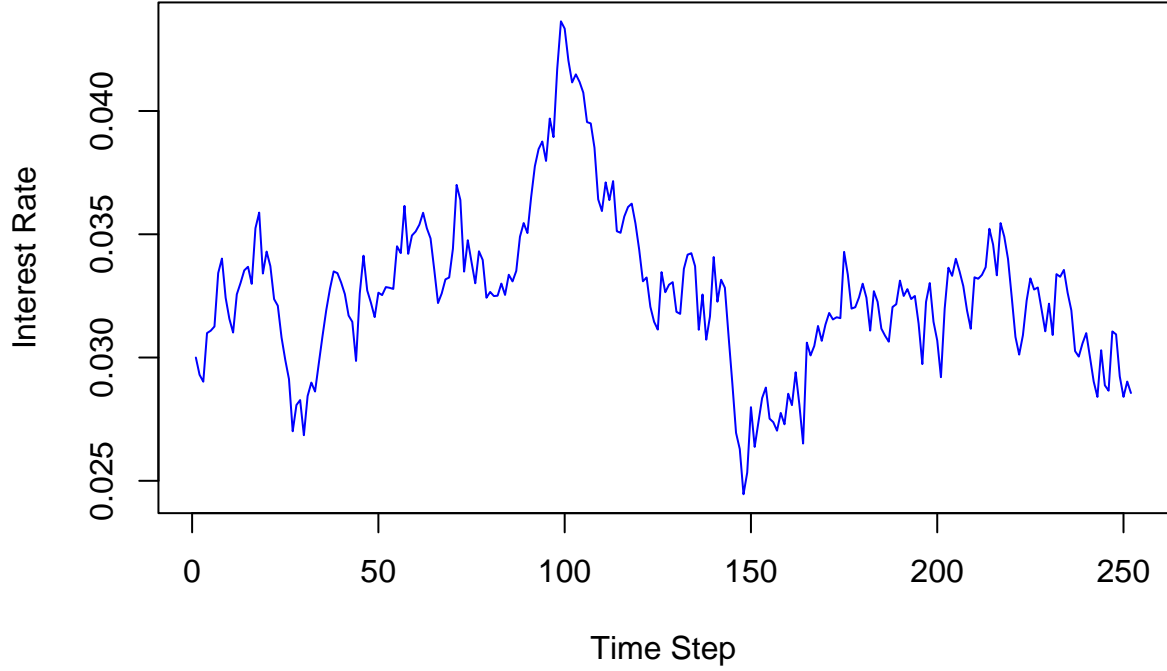
#Vasicek Process

set.seed(123)

for (i in 2:n) {
  # Weiner Process description of Brownian Motion
  dW <- rnorm(1, mean = 0, sd = sqrt(dt))
  # Passing process for n periods
  short_rate[i] <- short_rate[i-1] + a * (b-short_rate[i-1]) * dt + sigma * dW
}

plot(short_rate,
     type = "l",
     col = "blue",
     xlab = "Time Step",
     ylab = "Interest Rate",
     main = "Vasicek Model")
```

## Vasicek Model



## Cox-Ingersoll-Ross

### Derivation of the CIR Model

The Vasicek model is given by the following stochastic differential equation (SDE):

$$dr_t = \kappa(\theta - r_t)dt + \sigma dW_t, \quad (1)$$

where: -  $r_t$  is the short rate, -  $\kappa > 0$  is the speed of mean reversion, -  $\theta$  is the long-term mean, -  $\sigma$  is the volatility, -  $W_t$  is a standard Brownian motion.

While the Vasicek model allows negative interest rates, the Cox-Ingersoll-Ross (CIR) model modifies the diffusion term to avoid this.

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### 1. CIR Model Formulation

The CIR model introduces a square-root term in the volatility to ensure non-negativity of the short rate:

$$dr_t = \kappa(\theta - r_t)dt + \sigma\sqrt{r_t}dW_t. \quad (2)$$

Here, the volatility term  $\sigma\sqrt{r_t}$  ensures that as  $r_t \rightarrow 0$ , the stochastic term diminishes, preventing  $r_t$  from becoming negative.

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## 2. Mean-Reverting Property

Similar to the Vasicek model, the term  $\kappa(\theta - r_t)$  governs the mean-reverting drift. The CIR model is mean-reverting around  $\theta$  at a speed determined by  $\kappa$ .

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## 3. Solution of the CIR Model

To solve the CIR SDE, we first rewrite it in its general form:

$$dr_t = \kappa(\theta - r_t)dt + \sigma\sqrt{r_t}dW_t.$$

Using Itô's Lemma and the substitution  $r_t = x_t^2$ , we rewrite the CIR SDE for  $x_t$ :

$$dx_t = \left( \frac{\kappa\theta}{2x_t} - \frac{\kappa x_t}{2} \right) dt + \frac{\sigma}{2} dW_t. \quad (3)$$

This is a transformed SDE that can be solved numerically or analytically using advanced techniques.

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## 4. Feller Condition

The Feller condition ensures the non-negativity of  $r_t$ :

$$2\kappa\theta \geq \sigma^2. \quad (4)$$

This condition guarantees that the diffusion term  $\sigma\sqrt{r_t}$  does not dominate the drift term  $\kappa(\theta - r_t)$ , ensuring that  $r_t \geq 0$ .

---

## 5. Final Form

The CIR process can be expressed as:

$$r_t = r_0 e^{-\kappa t} + \theta \kappa \int_0^t e^{-\kappa(t-s)} ds + \sigma \int_0^t \sqrt{r_s} e^{-\kappa(t-s)} dW_s. \quad (5)$$

This solution highlights the mean-reverting nature, stochastic component, and the non-negativity feature of the CIR model.

```

# Cox-Ingersoll-Ross (CIR Model)
# Similar to Vasicek, bound above/at 0
# Implements additional process onto variance

# Define parameters for the CIR model
a <- 0.1      # Speed of mean reversion
b <- 0.05     # Long-term mean
sigma <- 0.02 # Volatility
r0 <- 0.03    # Initial interest rate
T <- 1       # Time horizon in years
dt <- 1/252  # Daily time step
n <- T/dt    # Number of steps

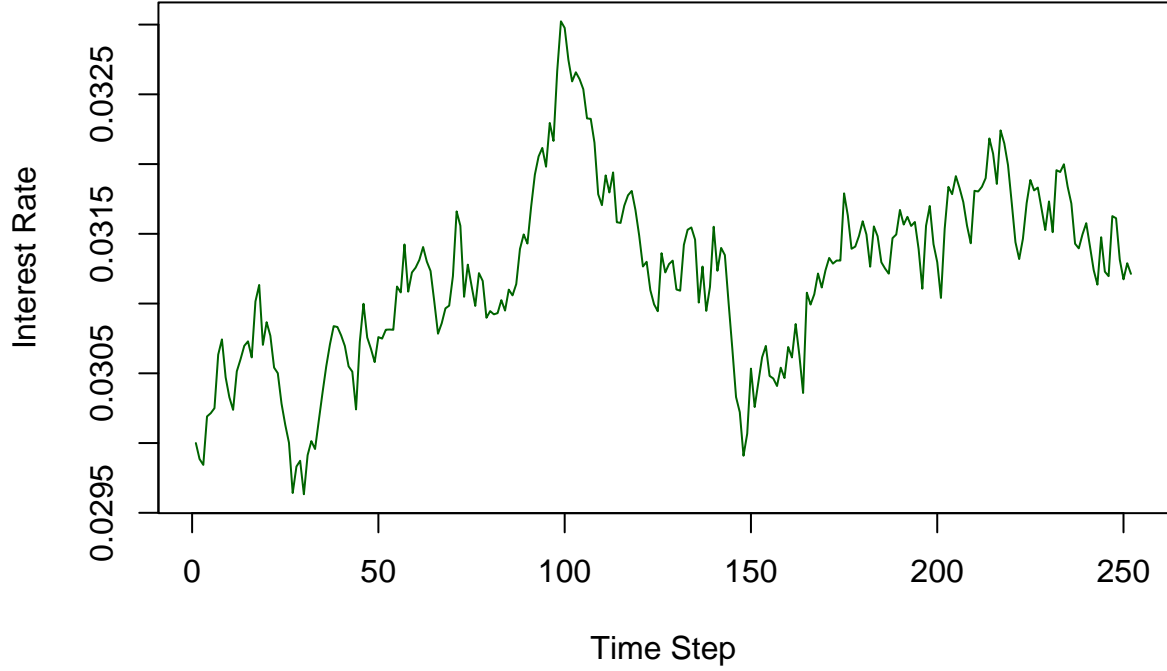
# Initialize short rate vector
short_rate <- numeric(n)
short_rate[1] <- r0

# Simulate CIR process
set.seed(123)
for (i in 2:n) {
  dW <- rnorm(1, mean = 0, sd = sqrt(dt))
  short_rate[i] <- short_rate[i-1] + a * (b - short_rate[i-1]) * dt + sigma * sqrt(short_rate[i-1]) * dW
  short_rate[i] <- max(short_rate[i], 0) # Ensure non-negativity
}

# Plot the simulated short rate path
plot(short_rate, type = "l", col = "darkgreen", xlab = "Time Step", ylab = "Interest Rate", main = "CIR")

```

## CIR Model Simulation



## Hull-White Model

### Derivation of the Hull-White Model

The Vasicek model is given by the following stochastic differential equation (SDE):

$$dr_t = \kappa(\theta - r_t)dt + \sigma dW_t, \quad (6)$$

where: -  $r_t$  is the short rate, -  $\kappa > 0$  is the speed of mean reversion, -  $\theta$  is the long-term mean, -  $\sigma$  is the volatility, -  $W_t$  is a standard Brownian motion.

While the Vasicek model assumes constant parameters, the **Hull-White model** introduces time-dependent drift to better fit the initial term structure of interest rates.

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### 1. Hull-White Model Formulation

The Hull-White model modifies the drift term in the Vasicek SDE to include a time-dependent mean  $\theta(t)$ :

$$dr_t = \kappa(\theta(t) - r_t)dt + \sigma dW_t. \quad (7)$$

Here: -  $\theta(t)$  is a deterministic function of time, often chosen to fit the initial term structure of interest rates.

---

## 2. Mean-Reverting Property

As with the Vasicek and CIR models, the term  $\kappa(\theta(t) - r_t)$  ensures mean reversion, but now the mean  $\theta(t)$  varies with time, allowing the model to adapt to the initial term structure.

---

## 3. Solution of the Hull-White SDE

To solve the SDE, we rewrite it:

$$dr_t + \kappa r_t dt = \kappa \theta(t) dt + \sigma dW_t.$$

Multiply through by the integrating factor  $e^{\kappa t}$ :

$$e^{\kappa t} dr_t + \kappa e^{\kappa t} r_t dt = \kappa \theta(t) e^{\kappa t} dt + \sigma e^{\kappa t} dW_t.$$

The left-hand side simplifies to an exact differential:

$$d(e^{\kappa t} r_t) = \kappa \theta(t) e^{\kappa t} dt + \sigma e^{\kappa t} dW_t.$$

Integrate both sides from 0 to  $t$ :

$$e^{\kappa t} r_t - r_0 = \int_0^t \kappa \theta(s) e^{\kappa s} ds + \int_0^t \sigma e^{\kappa s} dW_s. \quad (8)$$

---

## 4. Simplifying the Deterministic Term

The deterministic integral is:

$$\int_0^t \kappa \theta(s) e^{\kappa s} ds. \quad (9)$$

This integral depends on the functional form of  $\theta(t)$ . For specific choices of  $\theta(t)$ , such as a constant or linear function, it can be solved explicitly.

---

## 5. Stochastic Term

The stochastic integral remains as:

$$\int_0^t \sigma e^{\kappa s} dW_s. \quad (10)$$

The stochastic term contributes to the volatility of the short rate and can be expressed as a Wiener process with modified variance.

---



## 6. Final Solution

Combining the deterministic and stochastic components, the solution to the Hull-White model is:

$$r_t = r_0 e^{-\kappa t} + e^{-\kappa t} \int_0^t \kappa \theta(s) e^{\kappa s} ds + \sigma \int_0^t e^{-\kappa(t-s)} dW_s. \quad (11)$$

```
# Hull-White Model
```

```
#parameters
```

```
a <- 0.1
```

```
sigma <- 0.02
```

```
r0 <- 0.03
```

```
T <- 1
```

```
dt <- 1/252
```

```
n <- T/dt
```

```
short_rate <- numeric(n)
```

```
short_rate[1] <- r0
```

```
set.seed(123)
```

```
for (i in 2:n) {
```

```
  dW <- rnorm(1, mean = 0, sd = sqrt(dt))
```

```
  theta <- 0.05
```

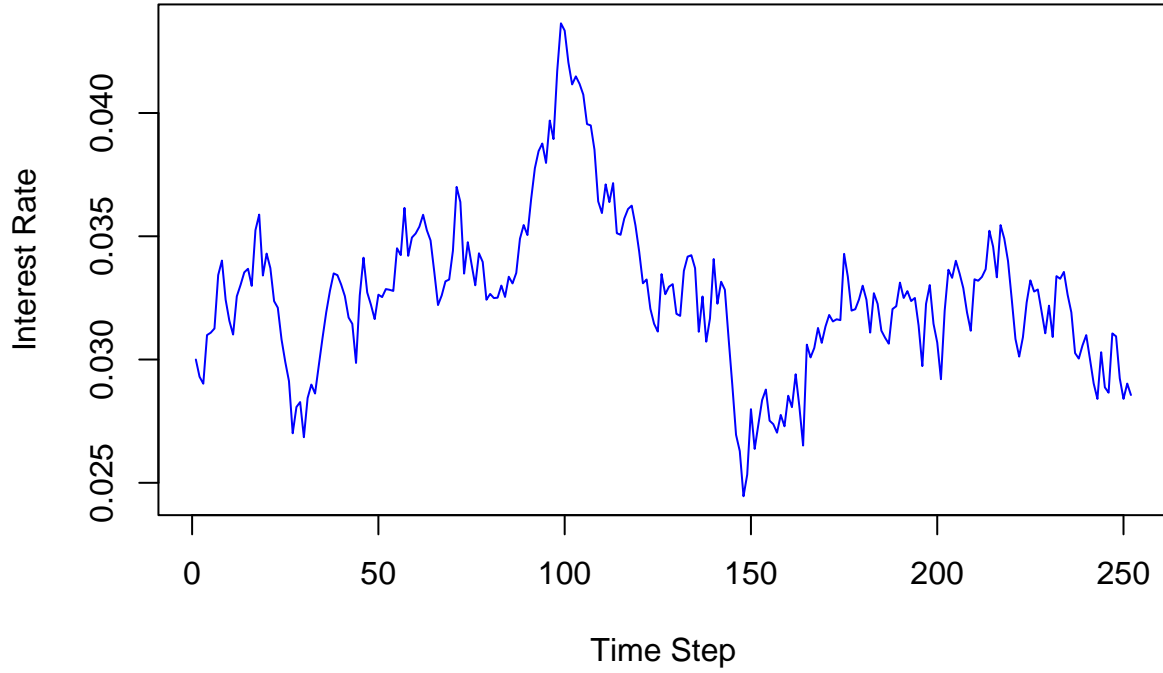
```
  short_rate[i] <- short_rate[i-1] + a * (theta - short_rate[i-1]) * dt + sigma * dW
```

```
}
```

```
# Plot the simulated short rate path
```

```
plot(short_rate, type = "l", col = "blue", xlab = "Time Step", ylab = "Interest Rate", main = "Hull-White")
```

## Hull-White Model Simulation



## Black - Karasinski Model

### Derivation of the Black-Karasinski Model

The Black-Karasinski model assumes that the logarithm of the short rate follows an Ornstein-Uhlenbeck process. Starting with the Vasicek model:

$$dr_t = \kappa(\theta - r_t)dt + \sigma dW_t,$$

where: -  $r_t$  is the short rate, -  $\kappa > 0$  is the mean reversion speed, -  $\theta$  is the long-term mean, -  $\sigma$  is the volatility, -  $W_t$  is a standard Brownian motion.

The Black-Karasinski model modifies this by defining the logarithm of the short rate,  $x_t = \log(r_t)$ , as following an Ornstein-Uhlenbeck process:

$$dx_t = \kappa(\mu(t) - x_t)dt + \sigma dW_t,$$

where: -  $x_t = \log(r_t)$ , -  $\mu(t)$  is a time-dependent mean function.

## 1. Model Formulation in Terms of $r_t$

Rewriting the SDE in terms of the short rate  $r_t = e^{x_t}$ , we have:

$$dr_t = r_t [\kappa(\mu(t) - \log(r_t))dt + \sigma dW_t].$$

This SDE ensures that  $r_t > 0$ , since the logarithmic transformation guarantees positivity of  $r_t$  throughout.

---

## 2. Mean-Reverting Property

The drift term  $r_t \kappa(\mu(t) - \log(r_t))dt$  ensures that the short rate reverts to a time-dependent mean  $e^{\mu(t)}$ . This mean-reverting property allows the model to flexibly adapt to changes in the term structure over time.

---

## 3. Solution of the Black-Karasinski Model

To solve the SDE, we first solve for  $x_t$ , which satisfies:

$$dx_t = \kappa(\mu(t) - x_t)dt + \sigma dW_t.$$

Multiply through by  $e^{\kappa t}$ , the integrating factor:

$$e^{\kappa t} dx_t + \kappa e^{\kappa t} x_t dt = \kappa \mu(t) e^{\kappa t} dt + \sigma e^{\kappa t} dW_t.$$

The left-hand side becomes:

$$d(e^{\kappa t} x_t) = \kappa \mu(t) e^{\kappa t} dt + \sigma e^{\kappa t} dW_t.$$

Integrating both sides from 0 to  $t$ , we get:

$$x_t = x_0 e^{-\kappa t} + \int_0^t \kappa \mu(s) e^{-\kappa(t-s)} ds + \sigma \int_0^t e^{-\kappa(t-s)} dW_s.$$

---

## 4. Back Transformation to $r_t$

Returning to the original short rate  $r_t = e^{x_t}$ , the solution becomes:

$$r_t = r_0^{e^{-\kappa t}} \exp \left( \int_0^t \kappa \mu(s) e^{-\kappa(t-s)} ds + \sigma \int_0^t e^{-\kappa(t-s)} dW_s \right).$$

This form highlights the log-normal distribution of  $r_t$  under the Black-Karasinski model.

```

# Define parameters for Black-Karasinski model
a <- 0.1      # Speed of mean reversion
sigma <- 0.02 # Volatility
r0 <- 0.03    # Initial interest rate
T <- 1        # Time horizon in years
dt <- 1/252   # Daily time step
n <- T/dt     # Number of steps

# Initialize short rate vector
short_rate <- numeric(n)
short_rate[1] <- log(r0)

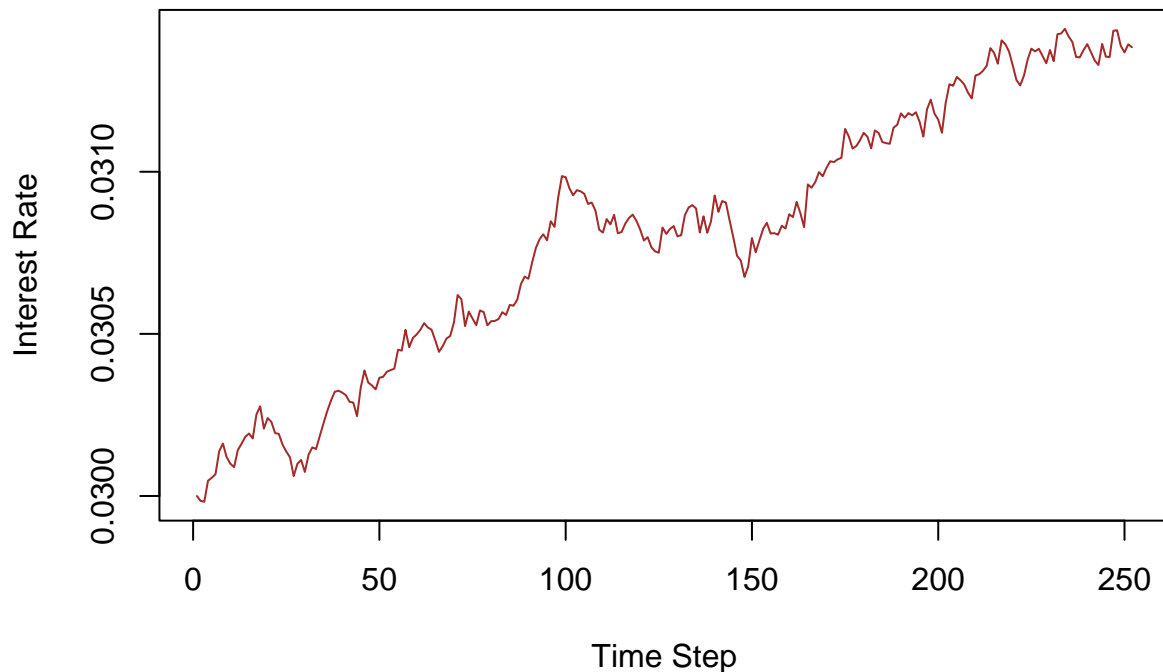
# Simulate Black-Karasinski process
set.seed(123)
for (i in 2:n) {
  dW <- rnorm(1, mean = 0, sd = sqrt(dt))
  short_rate[i] <- short_rate[i-1] + a * (log(0.05) - short_rate[i-1]) * dt + sigma * dW
}

# Convert back to interest rates
short_rate <- exp(short_rate)

# Plot the simulated short rate path
plot(short_rate, type = "l", col = "brown", xlab = "Time Step", ylab = "Interest Rate", main = "Black-K")

```

## Black-Karasinski Model Simulation



# Nelson - Siegel

## Derivation of the Nelson-Siegel Model

The **Nelson-Siegel model** is a parsimonious representation of the term structure of interest rates. It provides a functional form for the yield curve, characterized by level, slope, and curvature factors.

The yield at time  $t$  for a bond maturing at time  $T$  is expressed as:

$$y(t, \tau) = \beta_0 + \beta_1 \frac{1 - e^{-\lambda\tau}}{\lambda\tau} + \beta_2 \left( \frac{1 - e^{-\lambda\tau}}{\lambda\tau} - e^{-\lambda\tau} \right),$$

where: -  $y(t, \tau)$  is the yield at time  $t$  for maturity  $\tau = T - t$ , -  $\beta_0, \beta_1, \beta_2$  are parameters that represent level, slope, and curvature, -  $\lambda$  is a decay parameter that governs the exponential decay of the slope and curvature factors, -  $\tau$  is the time to maturity.

---

### 1. Model Intuition

The Nelson-Siegel model decomposes the yield curve into three components: 1. **Level** ( $\beta_0$ ): Represents the long-term average yield and shifts the entire curve up or down. 2. **Slope** ( $\beta_1$ ): Represents the short-term deviations from the level and controls the steepness of the curve. 3. **Curvature** ( $\beta_2$ ): Captures the hump-shaped behavior of the curve, typically around medium maturities.

The decay parameter  $\lambda$  determines the speed at which the slope and curvature factors decay to zero as maturity increases.

---

### 2. Functional Form

#### 2.1. Level Component ( $\beta_0$ )

The level factor is independent of  $\tau$ , making it constant across all maturities:

$$\text{Level: } \beta_0.$$

#### 2.2. Slope Component ( $\beta_1$ )

The slope factor decreases as  $\tau$  increases, with the decay controlled by  $\lambda$ :

$$\text{Slope: } \beta_1 \frac{1 - e^{-\lambda\tau}}{\lambda\tau}.$$

#### 2.3. Curvature Component ( $\beta_2$ )

The curvature factor increases initially, reaches a maximum, and then decays back to 0 for large  $\tau$ :

$$\text{Curvature: } \beta_2 \left( \frac{1 - e^{-\lambda\tau}}{\lambda\tau} - e^{-\lambda\tau} \right).$$

---

### 3. Special Cases

#### 3.1. When $\lambda \rightarrow 0$

The slope and curvature factors degenerate into linear terms, and the model reduces to a simple linear function of maturity.

#### 3.2. When $\lambda \rightarrow \infty$

The slope factor decays instantaneously, leaving only the level factor ( $\beta_0$ ) and curvature factor ( $\beta_2$ ) contributing to the yield.

---

### 4. Estimation

The parameters  $\beta_0, \beta_1, \beta_2$ , and  $\lambda$  are typically estimated using historical yield curve data by minimizing the squared error between the observed and modeled yields:

$$\min_{\beta_0, \beta_1, \beta_2, \lambda} \sum_{\tau} [y_{\text{obs}}(t, \tau) - y(t, \tau)]^2.$$

This estimation provides the best fit for the yield curve at a given time  $t$ .

---

### 5. Final Form

The Nelson-Siegel model can be summarized as:

$$y(t, \tau) = \beta_0 + \beta_1 \frac{1 - e^{-\lambda\tau}}{\lambda\tau} + \beta_2 \left( \frac{1 - e^{-\lambda\tau}}{\lambda\tau} - e^{-\lambda\tau} \right).$$

```
#Nelson-Siegel Model
library(minpack.lm)

# Define a function for Nelson-Siegel yield curve
nelson_siegel <- function(t, beta0, beta1, beta2, lambda) {
  return(beta0 + beta1 * (1 - exp(-lambda * t)) / (lambda * t) + beta2 * ((1 - exp(-lambda * t)) / (lambda * t) - exp(-lambda * t)))
}

# Example maturity and yield data
maturity <- c(1, 2, 3, 5, 7, 10)
yield <- c(0.02, 0.025, 0.027, 0.03, 0.032, 0.035)

# Fit the Nelson-Siegel model to the data
ns_model <- nlsLM(yield ~ nelson_siegel(maturity, beta0, beta1, beta2, lambda),
  start = list(beta0 = 0.03, beta1 = -0.01, beta2 = 0.01, lambda = 0.5))

# Display model summary
summary(ns_model)
```

```
##
## Formula: yield ~ nelson_siegel(maturity, beta0, beta1, beta2, lambda)
##
## Parameters:
##      Estimate Std. Error t value Pr(>|t|)
## beta0  3.901e-02  2.384e-03  16.36  0.00371 **
## beta1 -2.373e-02  1.908e-03 -12.44  0.00640 **
## beta2  6.005e-07  2.291e+00   0.00  1.00000
## lambda 5.133e-01  4.957e+01   0.01  0.99268
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.0009498 on 2 degrees of freedom
##
## Number of iterations to convergence: 14
## Achieved convergence tolerance: 1.49e-08
```

```
# Plot the fitted yield curve
fitted_yield <- predict(ns_model)
plot(maturity, yield, pch = 19, col = "red", xlab = "Maturity (Years)", ylab = "Yield", main = "Nelson-Siegel Yield Curve")
lines(maturity, fitted_yield, col = "blue", lwd = 2)
```

