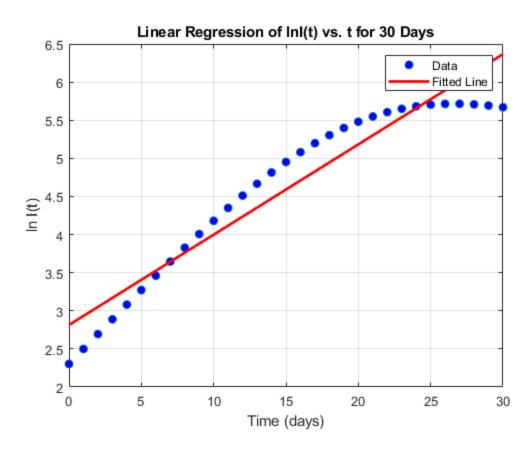
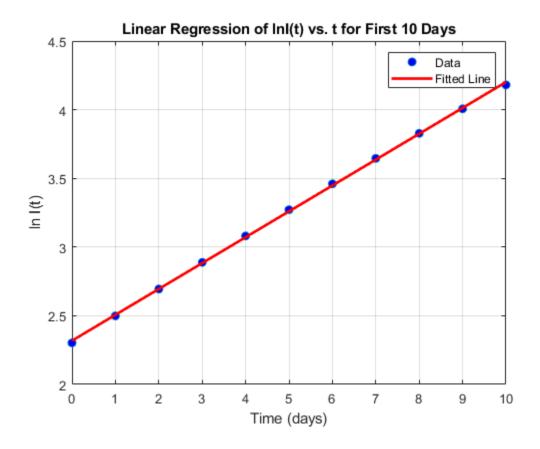
## **Part III: Linear Least Squares**

```
S(1) = 990;
I(1) = 10;
R(1) = 0;
h = 1;
Time = 30;
N total = 1000;
Beta = 0.3;
gamma = 0.1;
N = @(S, I, R) S + I + R;
dSdt = @(N pop, S curr, I curr) (-Beta / N pop) * S curr * I curr;
dIdt = @(N pop, S curr, I curr) (Beta / N pop) * S curr * I curr - gamma *
I curr;
dRdt = @(I curr) gamma * I curr;
Population(1) = S(1) + I(1) + R(1);
Population 2(1) = 0;
Population 3(1) = 0;
Population 4(1) = 0;
for i = 1:Time
    Population(i+1) = N(S(i), I(i), R(i));
    K1Susceptible = dSdt(Population(i), S(i), I(i));
    K1Infected = dIdt(Population(i), S(i), I(i));
    K1Recovered = dRdt(I(i));
   K2S = S(i) + 0.5 * K1Susceptible * h;
   K2I = I(i) + 0.5 * K1Infected * h;
    K2R = R(i) + 0.5 * K1Recovered * h;
    Population2(i+1) = N(K2S, K2I, K2R);
    K2Susceptible = dSdt(Population2(i+1), K2S, K2I);
    K2Infected = dIdt(Population2(i+1), K2S, K2I);
   K2Recovered = dRdt(K2I);
   K3S = S(i) + 0.5 * K2Susceptible * h;
    K3I = I(i) + 0.5 * K2Infected * h;
    K3R = R(i) + 0.5 * K2Recovered * h;
    Population3(i+1) = N(K3S, K3I, K3R);
    K3Susceptible = dSdt(Population3(i+1), K3S, K3I);
    K3Infected = dIdt(Population3(i+1), K3S, K3I);
    K3Recovered = dRdt(K3I);
   K4S = S(i) + K3Susceptible * h;
   K4I = I(i) + K3Infected * h;
   K4R = R(i) + K3Recovered * h;
    Population4(i+1) = N(K4S, K4I, K4R);
    K4Susceptible = dSdt(Population4(i+1), K4S, K4I);
    K4Infected = dIdt(Population4(i+1), K4S, K4I);
    K4Recovered = dRdt(K4I);
    S(i+1) = S(i) + (h / 6) * (K1Susceptible + 2 * K2Susceptible + 2 *
K3Susceptible + K4Susceptible);
    I(i+1) = I(i) + (h / 6) * (K1Infected + 2 * K2Infected + 2 * K3Infected
+ K4Infected);
```

```
R(i+1) = R(i) + (h / 6) * (K1Recovered + 2 * K2Recovered + 2 *
K3Recovered + K4Recovered);
end
t = 0:h:Time;
I t = I;
ln I t = log(I t);
n = length(t);
sum t = sum(t);
sum lnI = sum(ln I t);
sum t lnI = sum(t .* ln I t);
sum t squared = sum(t.^2);
numerator k = n * sum t lnI - sum t * sum lnI;
denominator k = n * sum t squared - sum t^2;
k = numerator k / denominator k;
ln I0 est = (sum lnI - k est * sum t) / n;
I0 est = exp(ln I0 est);
beta est = (k \text{ est + gamma}) * N \text{ total } / S(1);
fprintf('Using 30 days of data:\n');
fprintf('Estimated I(0) = %.5f, True I(0) = %.5f\n', I0 est, I(1));
fprintf('Estimated beta = %.5f, True beta = %.5f\n', beta est, Beta);
n 10 = 11; % Number of data points from t = 0 to t = 10
t 10 = t(1:n 10);
ln I t 10 = ln I t (1:n 10);
sum t 10 = sum(t 10);
sum lnI 10 = sum(ln I t 10);
sum t lnI 10 = sum(t 10 .* ln I t 10);
sum t squared 10 = sum(t 10.^2);
numerator k 10 = n 10 * sum t lnI 10 - sum t 10 * sum lnI 10;
denominator k 10 = n 10 * sum t squared 10 - sum t 10^2;
k est 10 = numerator k 10 / denominator k 10;
ln I0 est 10 = (sum lnI 10 - k est 10 * sum t 10) / n 10;
I0 est 10 = \exp(\ln 10 \text{ est } 10);
beta est 10 = (k \text{ est } 10 + \text{gamma}) * N \text{ total } / S(1);
fprintf('\nUsing first 10 days of data:\n');
fprintf('Estimated I(0) = %.5f, True I(0) = %.5f\n', I0 est 10, I(1));
fprintf('Estimated beta = %.5f, True beta = %.5f\n', beta est 10, Beta);
figure;
plot(t, ln I t, 'o', 'MarkerFaceColor', 'b');
hold on;
ln I fit = ln IO est + k est * t;
plot(t, ln I fit, 'r-', 'LineWidth', 2);
title('Linear Regression of lnI(t) vs. t for 30 Days');
```

```
xlabel('Time (days)');
ylabel('ln I(t)');
legend('Data', 'Fitted Line');
grid on;
hold off;
figure;
plot(t 10, ln I t 10, 'o', 'MarkerFaceColor', 'b');
hold on;
ln_I_fit_10 = ln_I0_est_10 + k_est_10 * t_10;
plot(t 10, ln I fit 10, 'r-', 'LineWidth', 2);
title('Linear Regression of lnI(t) vs. t for First 10 Days');
xlabel('Time (days)');
ylabel('ln I(t)');
legend('Data', 'Fitted Line');
grid on;
hold off;
Using 30 days of data:
Estimated I(0) = 16.67335, True I(0) = 10.00000
Estimated beta = 0.22065, True beta = 0.30000
Using first 10 days of data:
Estimated I(0) = 10.15965, True I(0) = 10.00000
Estimated beta = 0.29121, True beta = 0.30000
```





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Yes, using only the first 10 days of data provides estimates of I(0) and  $\beta$  closer to the true value. This is because the simplified model ( $S(t) \approx S_0$ ) holds better early on, making ln I(t) vs. t a more linear fit initially at least. From Part 1, we saw that the infected curve for seasonal influenza initially looks more linear, but over longer times, it appears more quadratic. Thus, shorter datasets give more accurate estimations for I(0) and  $\beta$ .