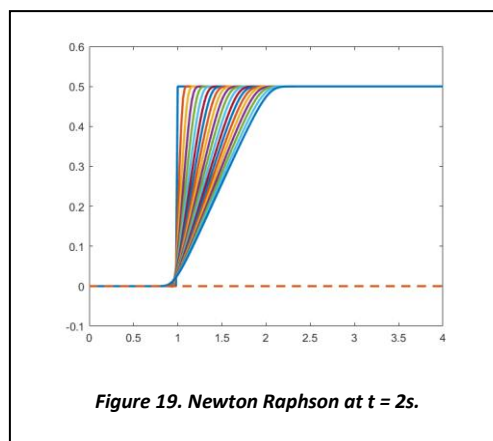
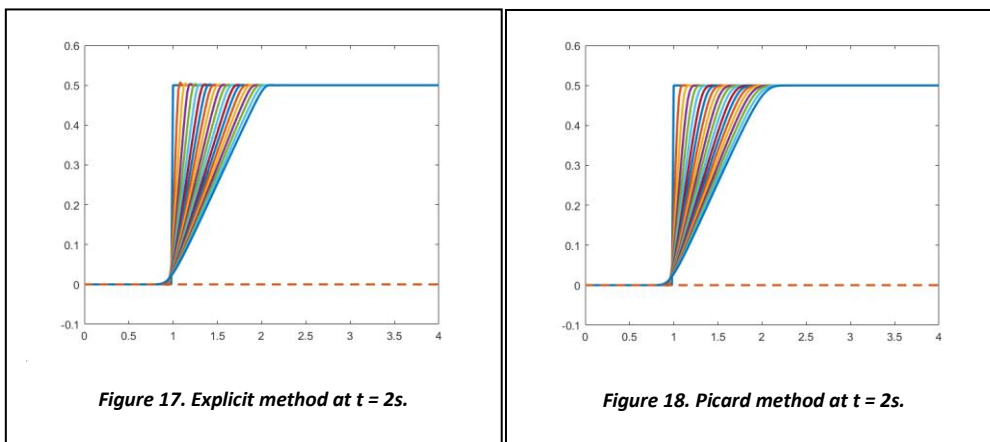
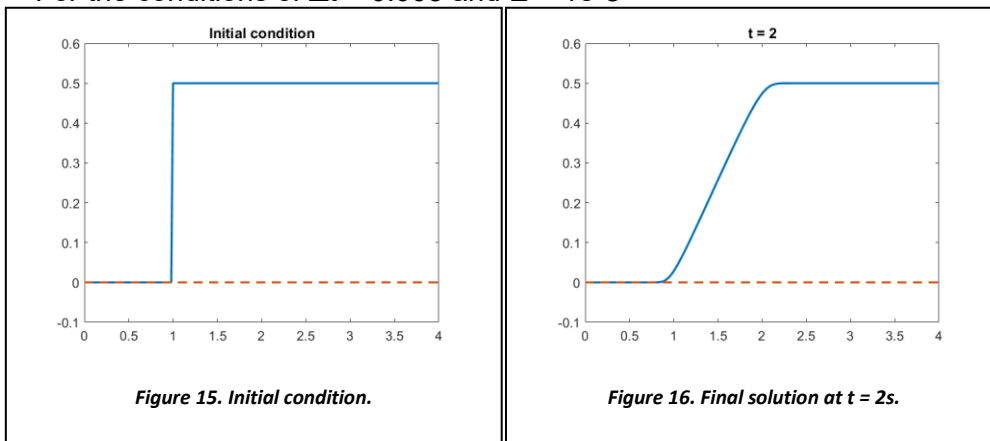


Problem statement

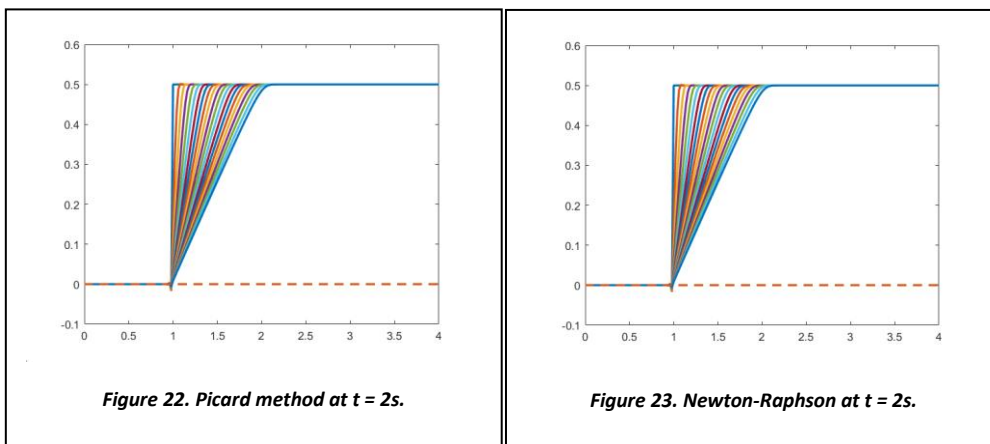
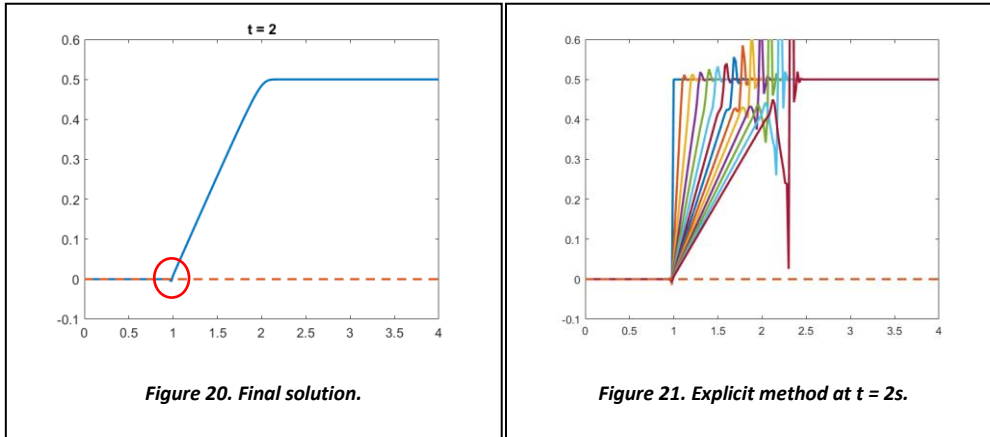
Case 1

```
u0 = [zeros(size(X(X<1))); 0.5*ones(size(X(X>=1)))];
```

- For the conditions of $\Delta t = 0.005$ and $E = 1e-3$



- For the conditions of $\Delta t = 0.005$ and $E = 0$



Comments on the results

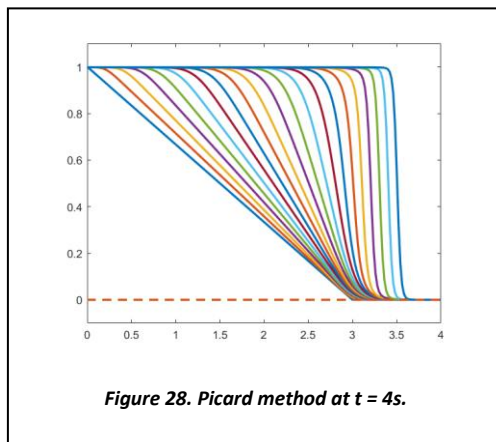
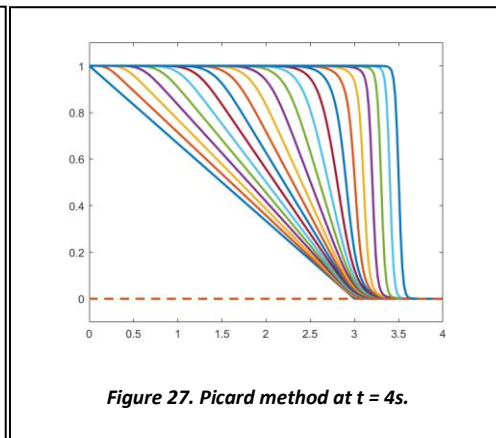
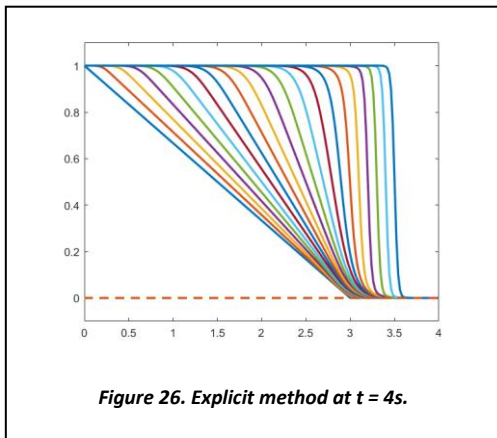
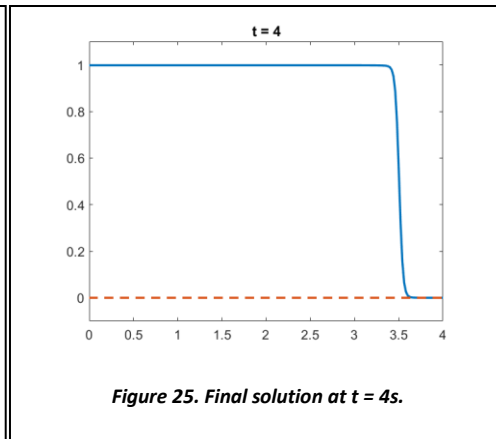
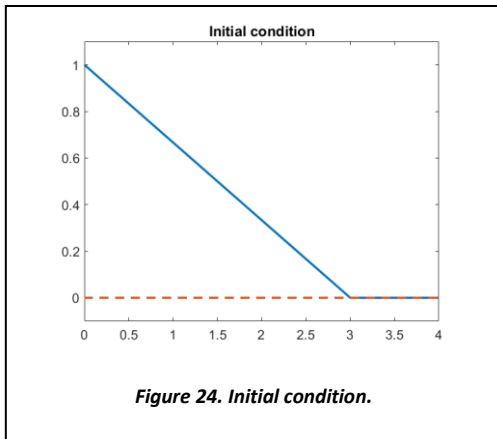
The main remark here is to point out the effect of having low timesteps and diffusivity coefficients closer to 0. As it is seen in the case of $\Delta t = 0.005$ and $E = 0$, it appears a lack of accuracy in the solution in implicit methods whereas for explicit methods the diffusivity seems not to be playing a role.

With increasing initial conditions, the solution gets the shape an entropy-compliant solution for Burger's equation (particularly rarefaction wave).

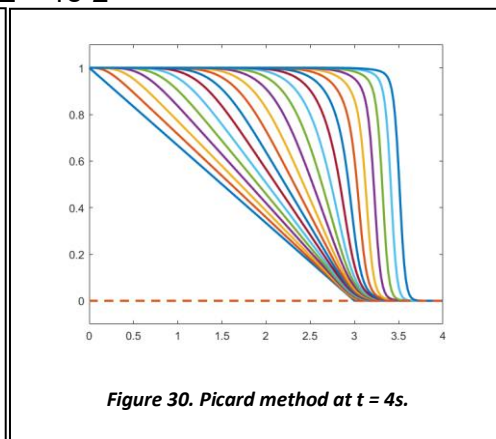
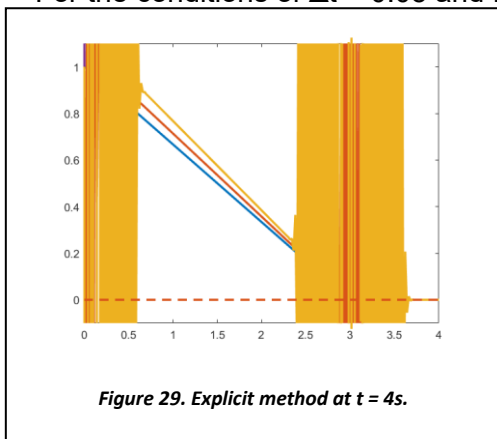
Case 2

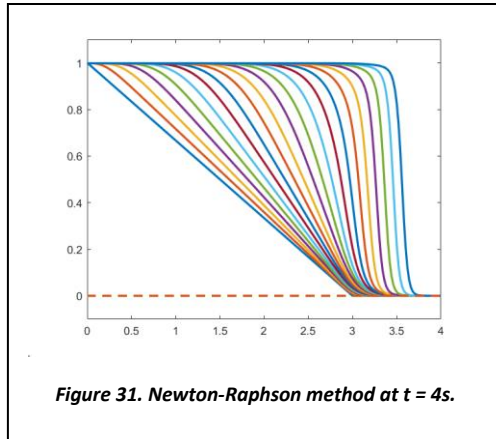
$$u_0 = [1 - x(x < 3) / 3; x(x \geq 3) * 0];$$

- For the conditions of $\Delta t = 0.005$ and $E = 1e-2$

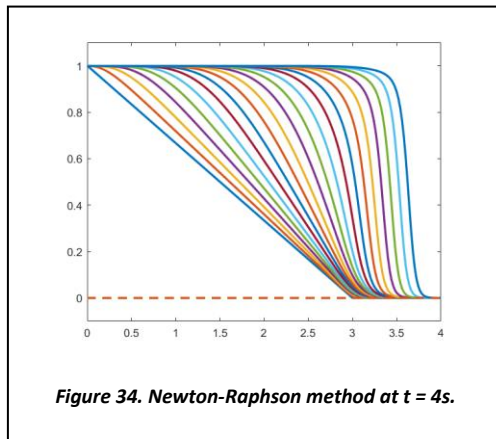
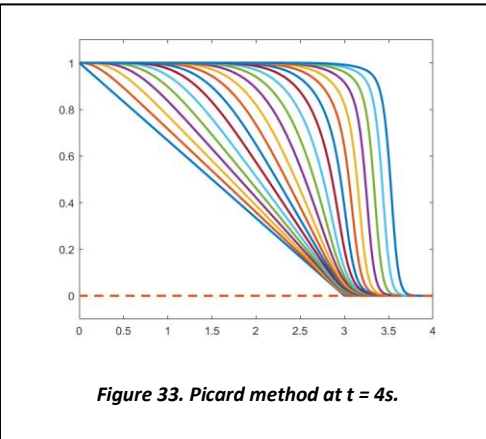
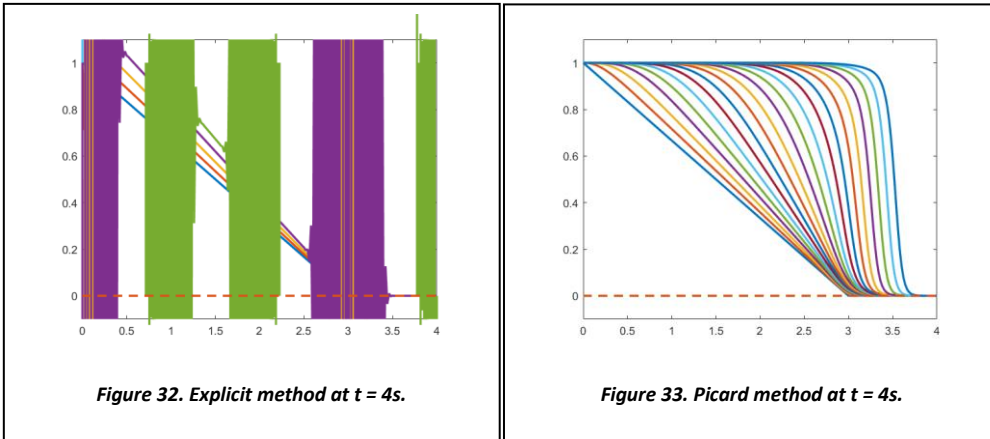


- For the conditions of $\Delta t = 0.05$ and $E = 1e-2$

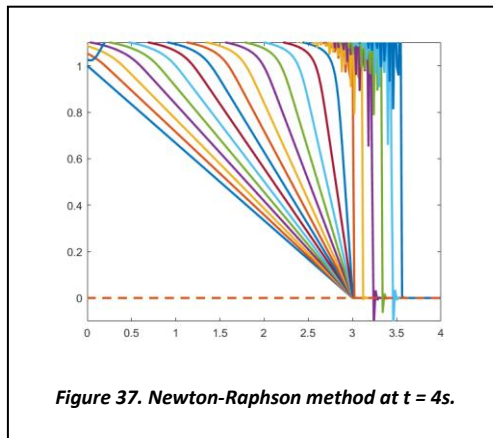
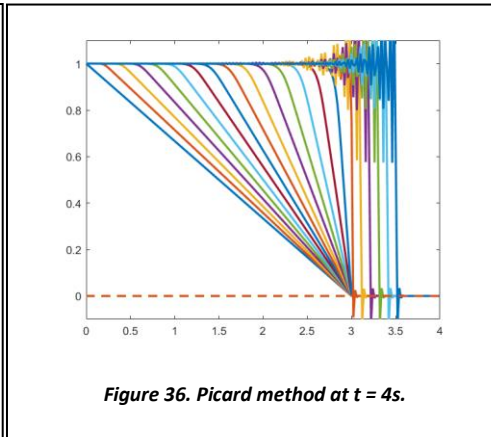
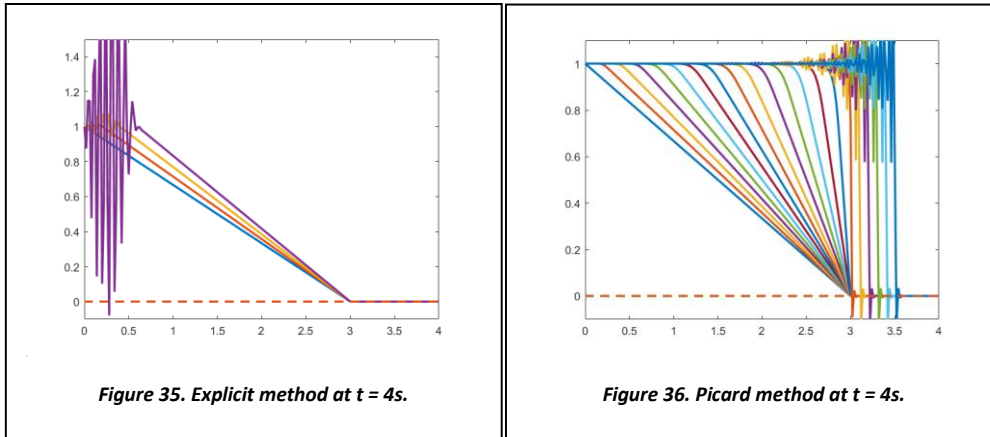




- For the conditions of $\Delta t = 0.1$ and $E = 1e-2$



- For the conditions of $\Delta t = 0.005$ and $E = 1e-4$



Comments on the results

For the case where we have initial decreasing data, it is got that explicit method shuts down when the time step is increased and the diffusion is lowered as it happens between cases $\Delta t = 0.005$ and $E = 1e-4$ and $\Delta t = 0.005$ and $E = 1e-2$. Also note that diffusivity helps the implicit methods, for decreasing initial data case, not to have discontinuities in the solution. Time step increment, when diffusivity is large enough, keep the solution of implicit methods stable.

Comments on the code

Code is mainly implemented following the steps in the slides. However, difficulties were faced when differentiating term $C(U)$.

```

for n = 1:nTimeSteps
    fprintf('\nTime step %d\n', n);
    U0 = U(:,n);
    error_U = 1; k = 0;

    while (error_U > 0.5e-5) && k < 20
        [C] = computeConvectionMatrix(X,T,U0);
        L = (M + At*C + At*E*K)*U0-M*U(:,n);
        P = M + 2*At*C+At*E*K;
        U1 = U0-P\L;
        error_U = norm(U1-U0)/norm(U1);
        fprintf('\t Iteration %d, error_U=%e\n',k,error_U);
        U0 = U1; k = k+1;
    end
    U(:,n+1) = U1;
end

```

First, L is computed as $M+At*C+At*E*K$ times the velocity vector U_0 which in this case takes the values that $U(:,n)$ has at every $nTimeStep$. So it is a vector that changes when the while condition is reached.

In the other hand, M is multiplying a matrix of $U(:,n)$, being n the number of time steps. This is a matrix that gets actualized also when the while condition is reached.

Once the condition is obtained, the U_0 vector, in the second iteration (p.e.) will take the value of last column solution of U. This means that last solution is used to compute the new one.

The jacobian in the Newton-Raphson method is the P value in the code. And the evaluation of the function F is L.

So the jacobian is implemented following the slides where the approach of the derivative of the convection matrix with respect to the solution is approximated by means of the residual method by 2 times $dt * C$.