

# Write-up: Project part 1

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## Pseudo-code for Conjugate Gradient (CG) algorithm

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### Algorithm 1 CG algorithm

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```
Initialize:  $u_0$   
 $r_0 = b - Au_0$   
 $p_0 = r_0$   
 $niter = 0$   
while  $niter < nitermax$  do  
     $niter = niter + 1$   
     $\alpha_n = r_n^T r_n / p_n^T A p_n$   
     $u_{n+1} = u_n + \alpha_n p_n$   
     $r_{n+1} = r_n - \alpha_n A p_n$   
    if  $\|r_{n+1}\|_2 / \|r_0\|_2 < threshold$  then  
        break  
    end if  
     $\beta_n = r_{n+1}^T r_{n+1} / r_n^T r_n$   
     $p_{n+1} = r_{n+1} + \beta_n p_n$   
end while
```

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### Answer:

The CG algorithm allows to solve  $Ax = b$  when  $A$  is sparse and with initial guess  $x$ . To avoid redundancy and use functions, I first wrote down the major steps of the algorithms, which were the following:

- Add or subtract vectors
- Multiply a vector by a scalar
- Dot product of two vectors
- 2-norm of a vector
- CSR matrix-vector multiplication

Based on this, I implemented those operations to build the CG algorithm using 5 different C++ functions, which are:

- `addVecsWithCoef(vec1, vec2, coef)`, which returns  $x + coef * y$
- `multiplyVecByScalar(vec, scalar)` to multiply `vec` by a scalar
- `dotPrd(x)`, which computes the dot product of two vectors
- `L2Norm(vec)`, which computes the L2 norm of a vector
- `csr_mat_vec_product(A, x)`, which computes the matrix-vector multiplication  $Ax$