Integral Operator Spectral Computations using PETSc/SLEPc libraries

Paulo B. Vasconcelos

Universidade do Porto, Portugal

Jose E. Roman

Universidad Politécnica de Valencia, Spain

Filomena Dias d'Almeida

Universidade do Porto, Portugal

Outline

- The Integral Problem
- Approximation and Discretization
- Iterative Refimement
- Multipower Double Iteration
- Implementation Using PETSc and SLEPc.
- Numerical tests
- Conclusions

Use of Parallel Libraries

- Illustrate
  - Advantages of parallel libraries
  - to support programming mathematical models
- Model
  - Integral operator on a very large interval
  - Radiative transfer in stellar atmospheres
  - Spectral computations $\leftrightarrow$ eigenvalues of matrix discretizations

The Integral Problem

- Radiative transfer in stellar atmospheres $Tx - x = f$,
- $T : X \to X$, where $X = L^1([0, t^\star])$

$$(Tx)(t) = \frac{\varpi}{2} \int_0^{t^\star} \left(\int_1^\infty \frac{\exp(-|t - t'|\mu)}{\mu} d\mu \right) x(t')dt', \quad t \in [0, t^\star].$$

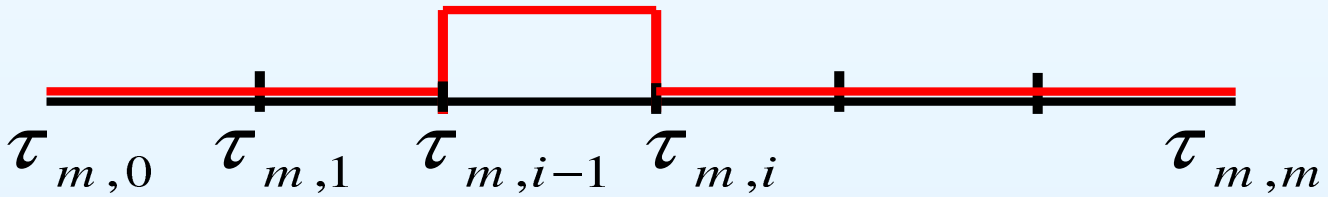
- $t^\star$ is the optical depth of the stellar atmosphere
- $\varpi \in]0, 1[$ is the albedo
- T has a singularity at the origin ($t = 0$).

Here we will compute the spectral elements (eigenvalues and eigenfunctions) of the operator T.

Approximation and Discretization

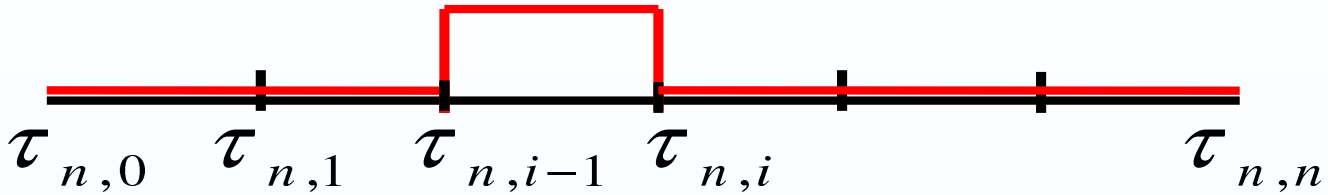
- Mathematical models for Scientific and Engineering problems need to be discretized for computational purposes into large dimensional matrices
- because either they do not have an analytic solution or it is too much complicated to deal with.

- The problem $T\varphi = \lambda\varphi$ defined in a infinite dimentional space X
- is approximated by $T_m\varphi_m = \lambda_m\varphi_m$ in X_m,
- m large enough.



Iterative Refimement

- To solve the approximating problem $T_m\varphi_m = \lambda_m\varphi_m$ in X_m.
- without solving $A_m x_m = \lambda_m x_m$,
- we use an iterative refinement strategy
- to improve the eigenvalues and vectors of a moderate size matrix A_n corresponding to a descretization in $X_n, n << m$.
- The refinement formula uses mainly matrix-vector multiplications with
- $A_m(m \times m)$ representing T in a very fine discretization space X_m.
- Large matrices are built and stored in a distributed manner among the processors,
- The iterative refinement provides accurate eigenpairs corresponding to a larger problem.



- The existence of the operator T linking the subspaces X_m and X_n allows the derivation of formulae to extend or restrict vectors from one of the finite dimensional subspaces to the other.
- The extension procedure will be denoted by E and the restriction by R

Initializations

- Given matrices A_n and A_m of dimensions n an m
- compute eigenpair of A_n (eigenvalue λ and eigenvector u) not necessarily the dominant, computed for instance Arnoldi's method
- The left eigenvector of A_n, say v, will be used as a normalization factor.

The procedure S returns y, given a vector w

- Solve the $(n+1) \times n$ linear system

$$\begin{cases} (A_n - \lambda I)t = Rw - (v^T R(w))u \\ v^T t = 0 \end{cases}$$

- compute the extension $y = E(A_n t)/\lambda$.

Multipower Double Iteration

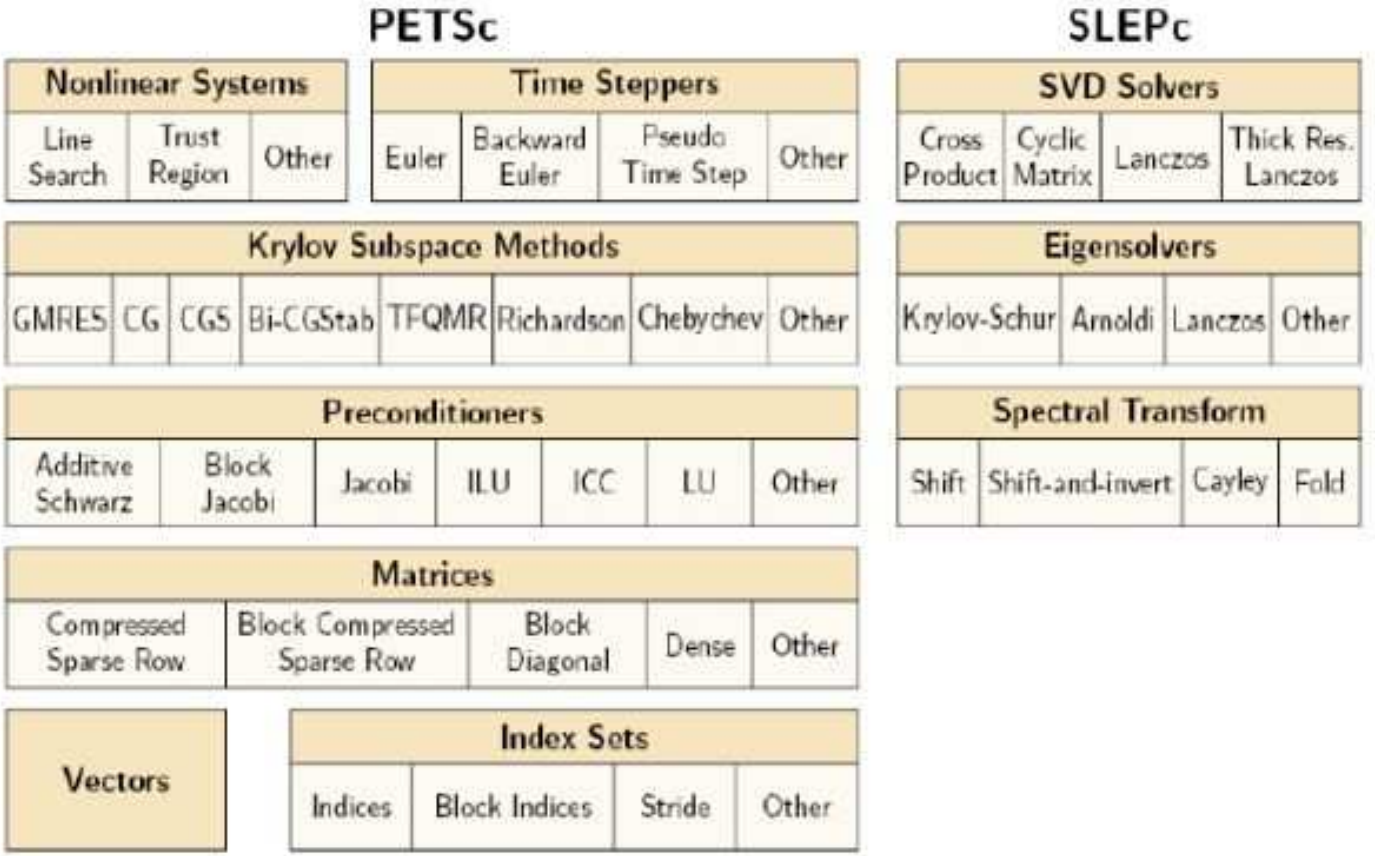
Algorithm:

1. Compute the eigenpair of A_n to be refined: λ, u and the left eigenvector v.
2. Extend eigenvalue: $x = E(u)$;
3. Compute the residual: $residual = norm(A_m x - \lambda x)$
4. While $residual$ less or equal a given tolerance
   (a) for j=1,2,. . . ,$itpower$
      i. $\mu = v^T R(A_m x)$
      ii. $x = A_m x / \mu$
   (b) $w = A_m x - \mu x$
   (c) Update refined eigenvector: $x = x - S(w)$
   (d) compute the new residual: $residual = norm(A_m x - \lambda x)$

PETSc/SLEPc

PETSc: Portable, Extensible Toolkit for Scientific Computation

SLEPc: Scalable Library for Eigenvalue Problem Computations



Libraries and the Algorithm

- Computation of initial approximations: eigenvalue and the corresponding right and left eigenvectors of matrix A_n
  - carried out only once with SLEPc, for instance with Arnoldi eigensolver
- Matrix-vector products with matrices A_n and A_m
  (also used in the restriction and extension from one grid to the other),
  - computed in a distributed way, are performed in PETSc with MatMult operation ;
    (all matrices are sparse and distributed)

Libraries and the Algorithm (cont)

- Vector operations,
  inner product, norm, addition and scaling,
  - are performed with the VecNorm, VecAXPY and VecScale routines from PETSc.
- Solution of linear systems (involved in procedure S)
  - are also carried out in PETSc.
  - Either by an LU decomposition of dimension $(n+1) \times n$ with partial pivoting.
    (This requires a modification of the corresponding routine in PETSC).
  - or by an iterative solver.

Libraries and the Algorithm (cont)

- Iterative solver.
  We use GMRES applied to $A_n - \lambda_n I$.
  - In order to get t satisfying $v^T t = 0$, we apply $I - uv^T$ to all Krylov vectors built by GMRES.
    (An implementation of the above scheme requires a modification of GMRES)
    - $(A_n - \lambda I)q$, the candidate vector to be added to the basis, is pre-multiplied by the projector $I - uv^T$.
    - Instead of $I - uv^T$, we consider the projector $I - vv^T$, since it can be implemented very easily in PETSc with function KSPSetNullSpace and it also guarantees $v^T t = 0$.
- PETSc provides many other iterative linear solvers and some preconditioners to improve the convergence .

Computer System Used

Odin cluster, located at Universidad Politécnica de Valencia.

- 55 dual-processor nodes
- 2.8 GHz Pentium Xeon processors , 1 GB of memory per node.
- interconnected with a high-speed SCI network with 2-D torus topology
- only one processor per node was used in the tests.

Numerical Tests

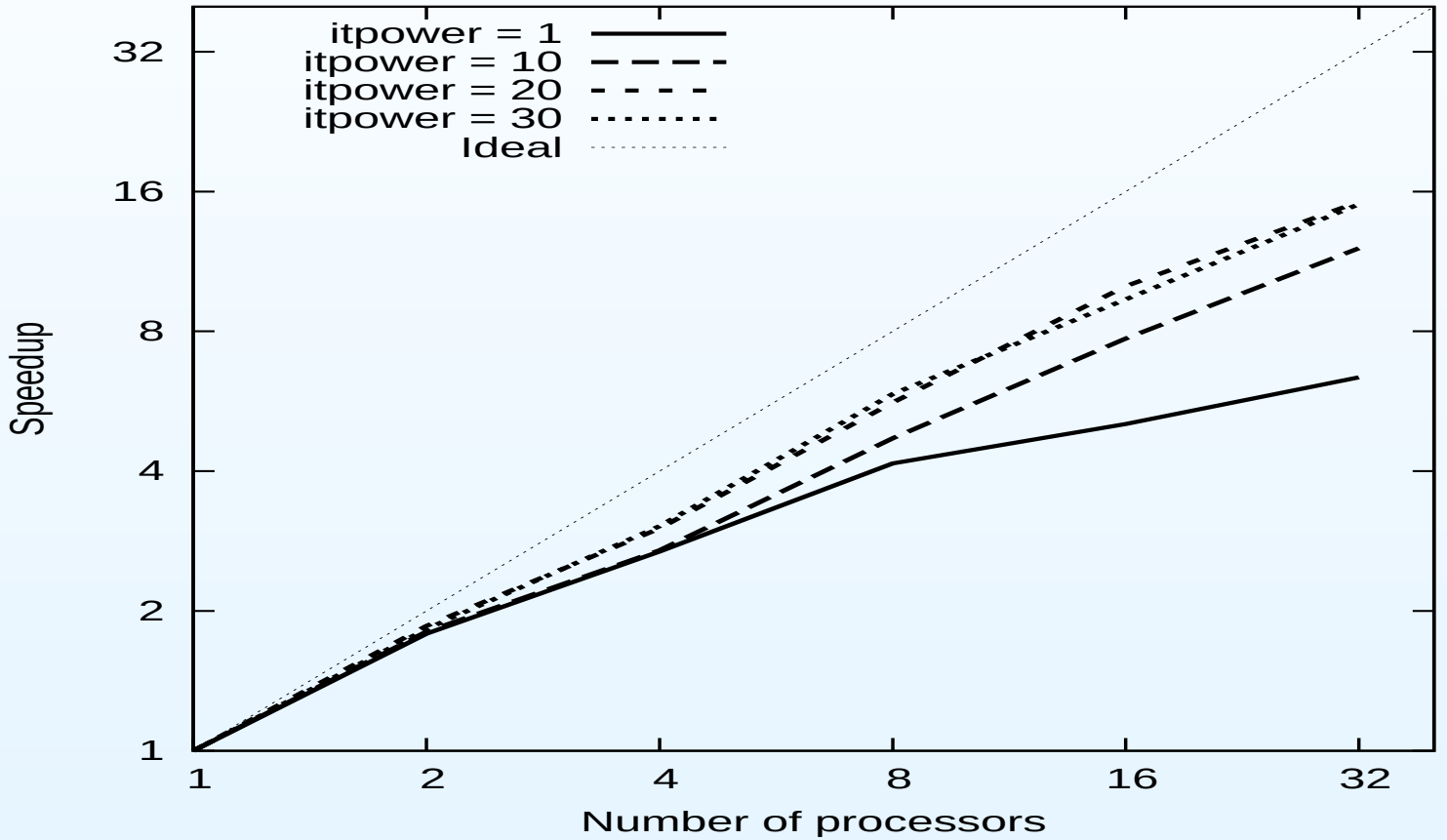
Performance analysis of the algorithm.

Several tests varying the number of processors and the algorithmic choices:

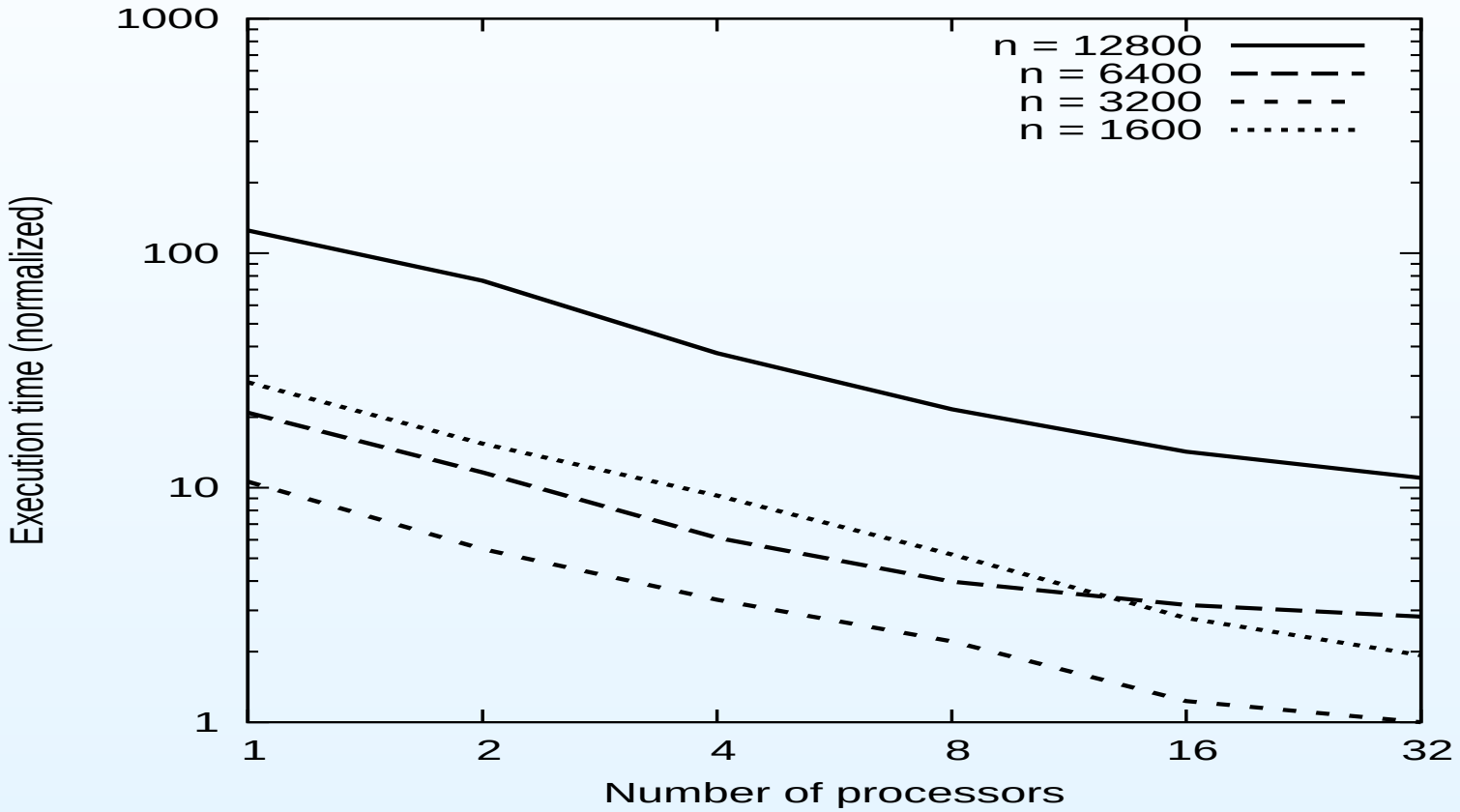
- influence of parameter $itpower$ (number of power iterations inside the refinement formula)
- influence of the dimension of initial problem
- effect of different preconditioners.

A tolerance of 10^{-12} on the residual

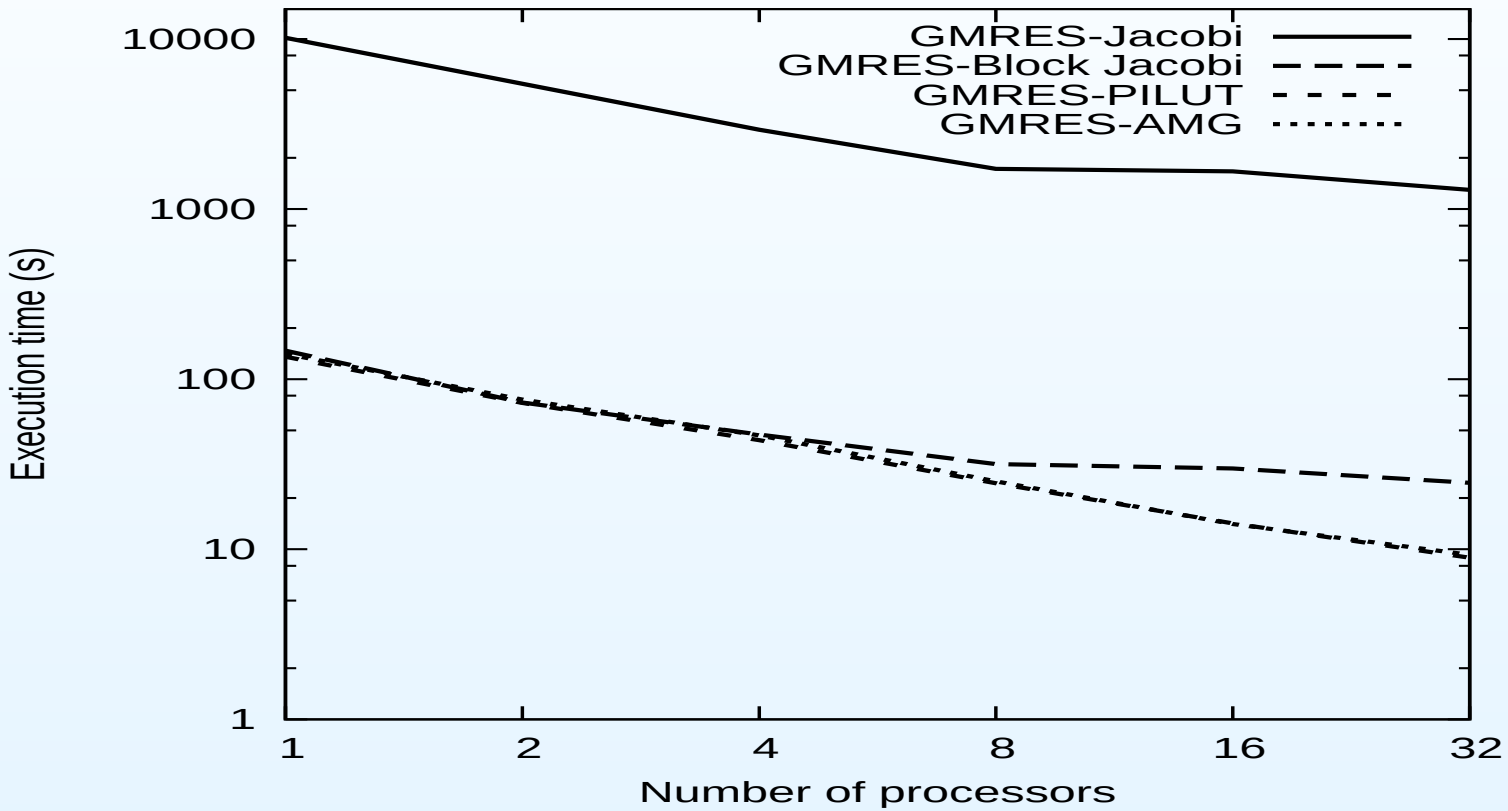
Speedup with different values of itpower



Times for a refined solution $m = 64000$, with several values of n



Several preconditioners to accelerate GMRES



Conclusions

- We presented a parallel code for the Multipower Defect Correction Method
  - refines eigenvalues/eigenfunctions from rough approximations
  - is effective for the computation of specific eigenvalues
- An implementation using libraries PETSc/SLEPc
  - shows good behaviour on a distributed parallel environment

References

[1] M. Ahues, F. D. d'Almeida, A. Largillier, P. B. Vasconcelos, "Defect correction for spectral computations for a singular integral operator", *Communications on Pure and Applied Analysis* **5(2)** pp. 241–250 (2006).

[2] F. D. d'Almeida and P. B. Vasconcelos, "Spectral Computations: from Operators to Matrices", *Proc. Appl. Math. Mech.* **7** 1022801–1022802 (2008).

[3] S. Balay, K. Buschelman, V. Eijkhout, W. D. Gropp, D. Kaushik, M. Knepley, L. C. McInnes, B. F. Smith, H. Zhang. "PETSc users manual", Tech. Rep. **ANL-95/11** - Revision 2.3.3, Argonne National Laboratory (2007).

[4] V. Hernandez, J. E. Roman, V. Vidal. "SLEPc: A scalable and flexible toolkit for the solution of eigenvalue problems", *ACM Transactions on Mathematical Software* **31 (3)** pp.351–362 (2005).

[5] V. Hernandez, J. E. Roman, A. Tomas, V. Vidal, "SLEPc users manual", Tech. Rep, **DSIC-II/24/02** - Revision 2.3.3", D. Sistemas Informáticos y Computación, Universidad Politécnica de Valencia (2007).