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# Minimal Problem Solver Generator

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<http://cmp.felk.cvut.cz/~trutmpav/theses/bsc-pavel-trutman.pdf>

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Text of acknowledgements. . .

## **Abstract**

Text of abstract...

## Resumé

Text of resumé...





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## Abbreviations

AHA! Some optional explanation before the list. Indentation can be set by the command `\setlength{\AbbrevIndent}{5em}`.

1D	one dimension(al)
2D, 3D, ...	two dimension(al), three dimension(al), two dimension(al), three dimension(al), two dimension(al), three dimension(al), two dimension(al), three dimension(al), ...
AAM	active appearance model
AI	artificial intelligence
ASM	active shape model
B-rep	boundary representation
BBN	Bayesian belief networks

# 1 Introduction

Here comes introduction.

## 2 Polynomial system solving

Firstly we review the state of the art algorithms for computing Gröbner basis. Better understanding of these algorithms helps us to integrate them more efficiently into polynomial solving algorithms based on Gröbner basis computation.

### 2.1 Buchberger's Algorithm

Buchberger's Algorithm [1], which was invented by Bruno Buchberger, was the first algorithm for computing Gröbner basis.

#### 2.1.1 First implementation

The first and easy, but very inefficient implementation of this algorithm is based on the observation that we can extend a set  $F$  of polynomials to a Gröbner basis only by adding all non-zero remainders  $\overline{S(f_i, f_j)}^F$  of all pairs from  $F$  into  $F$  until there is no non-zero remainder. The pseudocode of this algorithm can be found in [1] on page 87. The main disadvantage of this simple algorithm is that so constructed Gröbner basis are often bigger than necessary.

### 2.2 $F_4$ Algorithm

The  $F_4$  Algorithm [2] by Jean-Charles Faugère is an improved version of the Buchberger's Algorithm. The  $F_4$  replaces the classical polynomial reduction found in the Buchberger's Algorithm by a simultaneous reduction of several polynomials. This reduction mechanism is achieved by a symbolic precomputation followed by Gaussian elimination implemented using sparse linear algebra methods.  $F_4$  speeds up the reduction step by exchanging multiple polynomial divisions for row-reduction of a single matrix.

#### 2.2.1 Improved Algorithm $F_4$

The main function of  $F_4$  Algorithm consists of two parts. The goal of the first part is to initialize the whole algorithm. First, it generates required pairs and initializes Gröbner basis  $G$ . Then it takes each polynomial from the input set and calls function *Update* on it, which updates the set  $P$  of pairs and the set  $G$  of basic polynomials.

The second part of the algorithm generates new polynomials and includes them into the set  $G$ . In each iteration, it selects some pairs from  $P$  using function *Sel*. How to best select the pairs, is still an open question. Some selection strategies are described in the section 2.2.6 on page 6. Then, it splits each pair into two tuples. The first tuple contains the first polynomial  $f_1$  of the pair and monomial  $t_1$  such that  $\text{LM}(t_1 \times f_1) = \text{lcm}(\text{LM}(f_1), \text{LM}(f_2))$ . The second tuple is constructed in the same way from the second polynomial of the pair. All tuples from all selected pairs are put into the set  $L$ , i.e. duplicates are removed.

Next, it calls function *Reduction* on the set  $L$  and stores result in the set  $\tilde{F}^+$ . In the end it iterates through all new polynomials in the set  $\tilde{F}^+$  and calls function *Update* on each of them. This generates new pairs into the set  $P$  and extends Gröbner basis  $G$ .

This algorithm terminates when the set  $P$  of pairs is empty. Then the set  $G$  is a Gröbner basis and it is the output of the algorithm.

### 2.2.2 Function Update

In this algorithm is used standard implementation of Buchberger Criteria **CITE Gebauer and Moller [GM88]**.

### 2.2.3 Function Reduction

Task of this function is simple, it performs polynomial division using methods of linear algebra.

Input of this function is a set  $L$  containing tuples of monomial and polynomial, which were made in the main function of the  $F_4$  Algorithm.

First, this function calls function *Symbolic Preprocessing* on the set  $L$ . This returns a set  $F$  of polynomials to be reduced. To use linear algebra methods to perform polynomial division the polynomials have to be represented by a matrix. Each column of the matrix corresponds to a monomial and the columns have to be ordered with respect to the ordering used so that the most right column corresponds to “1”. Each row of the matrix corresponds to a polynomial from the set  $F$ . Construction of the matrix is simple. On the  $(i, j)$  position in the matrix, we put the coefficient of the term corresponding to  $j$ -th monomial from the  $i$ -th polynomial from the set  $F$ .

If we have constructed matrix like this we can reduce it to a row echelon form using, for example, Gauss-Jordan elimination. Note that this matrix is typically sparse so we can use sparse linear algebra methods to save computation time and memory. After elimination, we construct resulting polynomials by multiplying the reduced matrix by a vector of monomials from the ?????.

In the end, the function returns the set of reduced polynomials that have leading monomials which were not amongs polynomials before reduction.

### 2.2.4 Function Symbolic Preprocessing

Function *Symbolic Preprocessing* starts with a set  $L$  of tuples containing monomial and polynomial. These tuples were made from selected pairs. Then, these tuples are simplified by function *Simplify* and after multiplying polynomials with corresponding monomials the results are put into the set  $F$ .

Next, the function goes through all monomials in the set  $F$  and for each monomial  $m$  looks for some polynomial  $f$  from Gröbner basis  $G$  such  $m = m' \times \text{LM}(f)$  where  $m'$  is a some monomial. All such polynomials  $f$  and monomials  $m'$  are after simplification multiplied and put into the set  $F$ . The goal of this search is to have for each monomial in  $F$  some polynomial in  $F$  with the same leading monomial. This will ensure that all added polynomials will be reduced for  $G$  after polynomial division (using linear algebra).

### 2.2.5 Function Simplify

Function *Simplify* simplifies a polynomial which is a product of the multiplication of a given monomial  $m$  and a polynomial  $f$ .

The function recursively looks for a monomial  $m'$  and a polynomial  $f'$  such  $\text{LM}(t' \times f') = \text{LM}(t \times f)$ . The polynomial  $f'$  is selected from all polynomials that has been reduced in previous iterations (sets  $\tilde{F}^+$ ). We select polynomial  $f'$  such that total degree of  $m'$  is minimal.

This is done to insert into the set  $F$  (set of polynomials ready to reduce) polynomials such that are mostly reduced and have small number of monomials. This of course speeds up following reduction.

### 2.2.6 Selection strategy

For the speed of the  $F_4$  Algorithm is very important how to select in each iteration critical pairs from the list of all critical pairs  $P$ . This of course depends on the implemetation of the function  $Sel$ . There are more possible implemetations:

- The easiest implementation is to select all pairs from  $P$ . In this case we reduce all criticals pairs at the same time.
- If the function  $Sel$  selects only one critical pair then the  $F_4$  Algorithm is the Buchberger's Algorithm. In this case the  $Sel$  function corresponds to the selection strategy in the Buchberger's Algorithm.
- The best function that Faugère has tested is to select all critical pairs with a minimal total degree. Faugère calls this strategy the *normal strategy* for  $F_4$ .

## 2.3 $F_5$ Algorithm

## **3 Automatic generator**

### **3.1 Reimplementation**

### **3.2 Multiple eliminations solver**

### **3.3 Removing unnecessary polynomials**

### **3.4 Matrix partitioning**

### **3.5 F4 strategy**

## 4 Experiments



## 5 Conclusion

## Bibliography

- [1] David Cox, John Little, and Donald O'Shea. *Ideals, Varieties, and Algorithms : An Introduction to Computational Algebraic Geometry and Commutative Algebra*. Undergraduate Texts in Mathematics. Springer, New York, USA, 2nd edition, 1997. 4
- [2] Jean-Charles Faugère. A new efficient algorithm for computing gröbner bases ( $f_4$ ). *Journal of pure and applied algebra*, 139(1–3):61–88, 7 1999. 4