# Hybrid Symbolic-Numeric Methods in Polynomial Algebra 

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## Why hybrid symbolic-numeric methods?

In general, in polynomial algebra one is interested in solving problems whose input are polynomials with complex coefficients, i.e. the coefficients are only imperfectly known (i.e floating point numbers).

Example: Given $p(x)=x^{2}+1.99 x+1.00, q(x)=x+1.00$ and a tolerance $\delta=0.01$, compute the greatest common divisor of $p, q$, i.e. $\operatorname{gcd}(p, q)$ ! The tolerance $\delta=0.01$ means that the third and subsequent decimals of the coefficients are unknown!

In particular, we ${ }^{1}$ address a similar problem in polynomial algebra!

## Why hybrid symbolic-numeric methods?

Example: Given $f(x)=x^{2}-y^{3}-0.01 \in \mathbb{C}[x, y]$ squarefree, $\mathcal{C}=\left\{\left(x, y \in \mathbb{C}^{2} \mid f(x, y)=\right.\right.$ $0)\}$ plane complex algebraic curve and a tolerance $\delta=0.01$, compute a set of $\delta$-invariants of $\mathcal{C}$ (i.e. genus, etc) and its singularities (i.e. algebraic link, Alexander polynomial, etc).

We developed ${ }^{1}$ several symbolic-numeric algorithms for computing all these $\delta$-invariants. We presented them also in january at the Research Seminar in Rastenfeld.

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We implemented ${ }^{1}$ the algorithms in our library GENOM3CK using Axel. Support: http://people.ricam.oeaw.ac.at/m.hodorog/software.html ( M. Hodorog, B. Mourrain, J. Schicho. GENOM3CK - A library for GENus cOMputation of plane Complex algebraiC Curves using Knot theory. International Symposium on Symbolic and Algebraic Computation. Münich, Germany, 2010.


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Question: What does our algorithm really computes? What can we certify about the computed output? What do we mean by $\delta$-invariants?

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Question: What does our algorithm really computes? What can we certify about the computed output? What do we mean by $\delta$-invariants?
Answer: In order to provide our solution to these problems, we study different approaches for hybrid symbolic-numeric methods!

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## What are hybrid symbolic-numeric methods?

- We use the name "hybrid symbolic-numeric methods" as in the book "Computer Algebra Handbook" (Editors: J. Grabmeier, E. Kaltofen, V. Weispfenning).
- The objects of study are polynomials with both:
exact coefficients, i.e. integer and rational numbers: $1,-2, \frac{1}{2}$.
- and inexact coefficients, i.e. numerical values. For 1.001 we associate a tolerance of $10^{-3}$, i.e. the last digit is uncertain.


## What are hybrid symbolic-numeric methods?

- Numerical/approximate polynomial algebra

| Basic Notions |
| :---: |
| symbolic-numeric methods |
| approximate polynomials |
| ill-posed problems |

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Intuition. A symbolic-numeric method is similar to what Knuth calls a seminumerical algorithm, one that lies "on the borderline between numeric and symbolic computation."

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Remark. Consider $\mathbb{R}[x](\mathbb{C}[x])$ with the metric given by the euclidean norm $\|\cdot\|$. Given $f \in \mathbb{R}[x]$ and $\delta \in \mathbb{R}_{+}$, define an $\delta$-neighborhood of $f$ as:

$$
N_{f, \delta}=\{g \in \mathbb{R}[x]:\|f-g\| \leq \delta\} .
$$

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> Basic Notions
> symbolic-numeric methods approximate polynomials ill-posed problems

Definition. An ill-posed problem is a problem which does not fulfill Hadamard's definition of well-posedness:

- For all data, a solution exists.
- For all data, the solution is unique.
- The solution depends continously on the data (*).


## What are hybrid symbolic-numeric methods?

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| Basic Notions |
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Example 1 (ill-posed problem which does not fulfill (*)).
For $f(x)=x^{4}-1, g(x)=x^{2}+x-2$, get $\operatorname{gcd}(f, g)=x-1$.
For $\tilde{f}(x)=x^{4}-1-0.0001, \tilde{g}(x)=x^{2}+x-2-0.0001$, get $\operatorname{gcd}(\tilde{f}, \tilde{g})=1$.

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| Basic Notions |
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Example 2 (ill-posed problem which does not fulfill (*)).
Let $s=(0,0)$ singularity of $\mathcal{C}=\left\{(x, y) \in \mathbb{R}^{2} \mid-x^{3}-x y+y^{2}=0\right\}$, and $\mathcal{D}=\left\{(x, y) \in \mathbb{R}^{2} \mid-x^{3}-x y+y^{2}-0.01=0\right\}!$

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## Z. Zeng's approach

We distinguish between the theoretical part and the practical part. Theoretical part. Principles.

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- the set of problem instances with certain solution structure forms a pejorative manifold.
- For $P: D \rightarrow S$, we partition the set of input data $D$ in pejorative manifolds $M_{i}$ depending on the structure of the solution. $M_{i}$ form a stratification structure for $D$ !
- Tiny arbitrary perturbation pushes a problem instance away from its residing manifold, losing the structure of the solution.


## Z. Zeng's approach

Theoretical part. Example.

- Let $E: \mathbb{C}[x] \times \mathbb{C}[x] \rightarrow \mathbb{C}[x]$ an exact algorithm which Given $(f, g)$ assigns $E(f, g):=\operatorname{gcd}(f, g)$ !
This problem is ill-posed!
The stratification structure of $\mathbb{C}[x] \times \mathbb{C}[x]$ consists of:

$$
\left\{\begin{array}{l}
\mathcal{P}_{k}^{m, n}=\left\{(f, g) \in \mathbb{C}^{2}[x]: \operatorname{deg}(f)=m \geq \operatorname{deg}(g)=n, \operatorname{deg}(g c d(f, g))=k\right\} \\
\operatorname{codim} \mathcal{P}_{k}^{m, n}=k \in \mathbb{Z}_{+} \\
\overline{\mathcal{P}_{n}^{m, n}} \subset \overline{\mathcal{P}_{n-1}^{m, n}} \subset \ldots \subset \overline{\mathcal{P}_{0}^{m, n}}
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- Let $E: \mathbb{C}[x] \times \mathbb{C}[x] \rightarrow \mathbb{C}[x]$ an exact algorithm which Given $(f, g)$ assigns $E(f, g):=g c d(f, g)$ !
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$$

- $(f, g) \in \mathcal{P}_{k}^{m, n}: f=u v, g=u w, g c d(f, g)=u, \operatorname{deg}(u)=k, l c(u)=1$. Each $\mathcal{P}_{k}^{m, n}$ is parametrized as $F(\mathbf{u}, \mathbf{v}, \mathbf{w})=\mathbf{z}$, where $\mathbf{u}, \mathbf{v}, \mathbf{w}$, respectively $\mathbf{z}$ coefficients vectors of $u, v, w$, respectively $f, g$.


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- $f(x)=x^{4}-1, g(x)=x^{2}+x-2, g c d(f, g)=x-1$
- $\tilde{f}(x)=x^{4}-1.0001, \tilde{g}(x)=x^{2}+x-2.0001, \delta=10^{-4}, \operatorname{gcd}(\tilde{f}, \tilde{g})=1$.


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- $\mathbb{C}^{2}[x]$ has the stratification structure: $\overline{\mathcal{P}_{2}^{4,2}} \subset \overline{\mathcal{P}_{1}^{4,2}} \subset \overline{\mathcal{P}_{0}^{4,2}}$, where

$$
\mathcal{P}_{i}^{4,2}=\{(f, g): \operatorname{deg}(f)=4, \operatorname{deg}(g)=2, \operatorname{deg}(g c d(f, g))=i\}, i=0,1,2 .
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- $\operatorname{gcd}(f, g)=x-1 \in \mathcal{P}_{1}^{4,2}, \operatorname{gcd}(\tilde{f}, \tilde{g})=1 \in \mathcal{P}_{0}^{4,2}$.


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- $\operatorname{gcd}(f, g)=x-1 \in \mathcal{P}_{1}^{4,2}, \operatorname{gcd}(\tilde{f}, \tilde{g})=1 \in \mathcal{P}_{0}^{4,2}$.
- Problem: Given $\tilde{f}, \tilde{g}, \delta$ and not knowing $f, g$, identify $\mathcal{P}_{1}^{4,2}$ !


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- $\operatorname{gcd}(f, g)=x-1 \in \mathcal{P}_{1}^{4,2}, \operatorname{gcd}(\tilde{f}, \tilde{g})=1 \in \mathcal{P}_{0}^{4,2}$.
- Problem: Given $\tilde{f}, \tilde{g}, \delta$ and not knowing $f, g$, identify $\mathcal{P}_{1}^{4,2}$ !
- Answer (Z. Zeng): $\mathcal{P}_{1}^{4,2}$ is the highest codimension manifold among all manifolds that intersect the $\delta$-neighborhood of $\tilde{f}, \tilde{g}$ !


## Z. Zeng's approach

Theoretical part. Principles.
Consider $\tilde{P}$ a perturbation from an exact problem $P$ with sufficiently small error
$\delta$. Formulate an approximate solution to $\tilde{P}$ using the 3 -strikes principle:

- The approximate solution of $\tilde{P}$ is the exact solution of a problem $\hat{P}$ within $\delta$.
- $\hat{P}$ is on the highest codimension manifold $\Pi$ intersecting the $N_{\tilde{P}, \delta}$.
- $\hat{P}$ is the nearest problem to $\tilde{P}$ on $\Pi$.

Remark: This approximate solution satisfy the property: as the error $\delta$ approaches 0 the approximate solution converges to the exact solution.

## Z. Zeng's approach

Practical part. A 2-stage approach may help to solve an ill-posed problem:

- Stage 1: maximizing the codimension of the manifolds.
(i.e. determine the structure of the solution).

Computation tools: matrix building.

- Stage 2: minimizing the distance to the manifold.

Computation tools: nonlinear least-squares, Gauss-Newton iteration.

## Z. Zeng's approach

Practical part. Example.
Given $(\tilde{f}, \tilde{g}, \delta) \in \mathbb{C}^{2}[x] \times \mathbb{R}_{+}, m=\operatorname{deg}(f) \geq n=\operatorname{deg}(g)$ compute the $g c d(\tilde{f}, \tilde{g})$ !

- Stage 1: Compute the degree of $\operatorname{gcd}(\tilde{f}, \tilde{g})$ !
- Stage 2: Compute the coefficients of $\operatorname{gcd}(\tilde{f}, \tilde{g})$ !


## Z. Zeng's approach

Practical part. Example.
Stage 1. Compute the degree of $\operatorname{gcd}(\tilde{f}, \tilde{g})$ ! by using a low rank approximation of the Sylvester matrix of $(\tilde{f}, \tilde{g})$, i.e. $S:=S(\tilde{f}, \tilde{g})$ !

- Theorem 1: $\operatorname{deg}(\operatorname{gcd}(\tilde{f}, \tilde{g}))=m+n-\operatorname{rank}(S(\tilde{f}, \tilde{g}))$.


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- How to compute $\operatorname{rank}(S(\tilde{f}, \tilde{g}))$ in the presence of data perturbations?
- Intuition: If $S$ is rank deficient, then small perturbations of the matrix values can yield a matrix of full rank! We approximate $S$ by a low rank matrix $\tilde{S}$, by Singular Value Decomposition (SVD) of $S$.


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- Intuition: If $S$ is rank deficient, then small perturbations of the matrix values can yield a matrix of full rank! We approximate $S$ by a low rank matrix $\tilde{S}$, by Singular Value Decomposition (SVD) of $S$.
- Theorem 2: Let $A \in \mathcal{M}_{m \times n}(\mathbb{C})$ with $m \geq n$. Then there exists $U \in \mathcal{M}_{m \times m}, V \in \mathcal{M}_{n \times n}$ orthogonal, and a unique $\Sigma(A) \in \mathcal{M}_{m \times n}$ with $\Sigma(A)=\operatorname{diag}\left(\sigma_{1} \geq \ldots \geq \sigma_{r}\right)$ s.t. $A=U \Sigma V^{t}$.
- Remark: $\operatorname{rank}(A)=r$ and $r \leq \min (m, n)=n$.


## Z. Zeng's approach

Practical part. Example.
Stage 1. Compute the degree of $\operatorname{gcd}(\tilde{f}, \tilde{g})$ !

- Theorem 3: Let $S \in \mathcal{M}_{m \times n}$ with $\Sigma(S)=\operatorname{diag}\left(\sigma_{1}, \sigma_{2}, \ldots, \sigma_{n}\right)$ as in Theorem 2. Assume $\sigma_{1} \geq \sigma_{2} \geq \ldots \geq \sigma_{k}>\theta \geq \sigma_{k+1} \geq \ldots \geq \sigma_{n}$ for $\theta \in \mathbb{R}_{+}$. Then there exists $\tilde{S}$, with $\Sigma(\tilde{S})=\operatorname{diag}\left(\sigma_{1}, \sigma_{2}, \ldots, \sigma_{k}\right)$, i.e. $\operatorname{rank}(\tilde{S})=k$ and

$$
\min _{\operatorname{rank}(B)=k}\|S-B\|=\|S-\tilde{S}\| \leq \theta .
$$

- Remark: By dropping insignificant singular values of $S$ (i.e. all $\sigma_{i} \leq \theta$ ) we obtain $\tilde{S}$ with $\operatorname{rank}(\tilde{S})<\operatorname{rank}(S)$ and $\|S-\tilde{S}\| \leq \theta$ !


## Z. Zeng's approach

Practical part. Example.
Stage 2. Compute the coefficients of $\operatorname{gcd}(\tilde{f}, \tilde{g})$ with $\operatorname{deg}=k$ !

- Find $(\hat{u}, \hat{v}, \hat{w})$ with $\operatorname{gcd}(\tilde{f}, \tilde{g}) \cong \hat{u}, \operatorname{gcd}(\hat{v}, \hat{w})=1, \operatorname{deg}(\hat{u})=k$ and

$$
\left\{\begin{array}{l}
\hat{u} \hat{v} \cong \tilde{f}  \tag{2}\\
\hat{u} \hat{w} \cong \tilde{g}
\end{array}\right.
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\hat{u} \hat{v} \cong \tilde{f}  \tag{2}\\
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$$

- Rewrite (2) as $F(\mathbf{u}, \mathbf{v}, \mathbf{w}) \cong \mathbf{b}$, where $\mathbf{u}, \mathbf{v}, \mathbf{w}$, respectively $\mathbf{b}$ represents the coefficients vectors of the polynomials $\hat{u}, \hat{v}, \hat{w}$, respectively $\tilde{f}, \tilde{g}$.


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Stage 2. Compute the coefficients of $\operatorname{gcd}(\tilde{f}, \tilde{g})$ with $\operatorname{deg}=k$ !

- Find $(\hat{u}, \hat{v}, \hat{w})$ with $\operatorname{gcd}(\tilde{f}, \tilde{g}) \cong \hat{u}, \operatorname{gcd}(\hat{v}, \hat{w})=1, \operatorname{deg}(\hat{u})=k$ and

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- Solve overdeterminate system $F(\mathbf{u}, \mathbf{v}, \mathbf{w}) \cong \mathbf{b}$. Solve $\min _{(u, v, w) \in \mathcal{P}_{k}^{m, n}}\|F(u, v, w)-b\|=\|F(\hat{u}, \hat{v}, \hat{w})-b\|$ by Gauss-Newton.
- Necessary: the coefficients of the $g c d$ must be real numbers! When the coefficients are integers, Stage 2 cannot be used (our case)!
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## E. Kaltofen's approach

## PROBLEM

- Given: $f, g \in \mathbb{C}[x], \operatorname{deg}(f)=m, \operatorname{deg}(g)=n, k \leq \min (m, n) \in \mathbb{Z}_{+}$


## E. Kaltofen's approach

## PROBLEM

- Given: $f, g \in \mathbb{C}[x], \operatorname{deg}(f)=m, \operatorname{deg}(g)=n, k \leq \min (m, n) \in \mathbb{Z}_{+}$
- find: $\tilde{f}, \tilde{g} \in \mathbb{C}[x]$
- s.t. $\min _{\tilde{\tilde{f}} \tilde{\tilde{q}}) \geq h}\|\tilde{f}-f\|^{2}+\|\tilde{g}-g\|^{2}:=\mathcal{N}$, and $\operatorname{deg}(\operatorname{gcd}(\tilde{f}, \tilde{g})) \geq k$
$\operatorname{deg}(\|\tilde{f}-f\|) \leq m, \operatorname{deg}(\|\tilde{g}-g\| \leq n)$.


## E. Kaltofen's approach

## METHOD

- Prove the existence of $\mathcal{N}$.
- Compute $\tilde{f}, \tilde{g}, \mathcal{N}$ by an iterative algorithm denoted IterativeAlgo given $f, g, k$, tol $\in \mathbb{R}_{+}$, based on:


## Theorem 4

Given $f(x), g(x) \in \mathbb{C}[x]$ with $\operatorname{deg}(f)=m, \operatorname{deg}(g)=n$. Let $S(f, g)$ the Sylvester matrix of $f, g$ and $S_{k}$ the $k$-th Sylvester matrix, $1 \leq k \leq \min (m, n)$. Then:

$$
\operatorname{deg}(g c d(f, g)) \geq k \Leftrightarrow \operatorname{rank}(S) \leq m+n-k \Leftrightarrow \operatorname{dim} K e r\left(S_{k}\right) \geq 1
$$

Let $S_{k}=\left[b_{k} A_{k}\right], b_{k}$ is the first column of $S_{k}, A_{k}$ the remaining columns. Then $\operatorname{dim} \operatorname{Ker}\left(S_{k}\right) \geq 1 \Leftrightarrow A_{k} x=b_{k}$ has a solution.

## E. Kaltofen's approach

## APPLICATION

Algorithm 1 Approximate gcd of univariate polynomials
Input: $f, g \in \mathbb{C}[x], \operatorname{deg}(f)=m \geq n=\operatorname{deg}(g), \epsilon \in \mathbb{R}_{+}$
Output: $\epsilon-g c d(f, g)$

- Initialize $k=n$
- Repeat
- Compute $\tilde{f}, \tilde{g}, \mathcal{N}$ with IterativeAlgo( $f, g, k, t o l)$.
- $k$ - -
- until $\mathcal{N}<\epsilon$ or $k<0$
- If $k \geq 0$ then compute $\epsilon$ - gcd from the matrix $S_{k}(\tilde{f}, \tilde{g})$, for instance with an algorithm like Zeng's algorithm based on SVD.


## (1) Why hybrid symbolic-numeric methods?

(2) What are hybrid symbolic-numeric methods?
(3) Different approaches for hybrid symbolic-numeric methods

- Z. Zeng's (H. J. Stetter's) approach
- E. Kaltofen's approach
- C. Yap's approach
- Shortly about our approach


## C. Yap's approach (exact geometric computation)

Constructive zero bounds

- an expression $E$ is a syntactic object constructed from a set of operators $\Omega$ over $\mathbb{R}$.
- evaluating predicates amounts to determining the sign of $E$.


## Definition

$b>0$ is a zero bound for $E$ if the following holds: if $E$ is well-defined $(E \neq \uparrow)$ and $E \neq 0$ then $|E| \geq b .-\log _{2}(b)$ is a zero bit-bound for $E$.

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- Given $E$ determine $\operatorname{sign}(E)=\{\uparrow,+1,-1,0\}$ !
- If $E \neq \uparrow$ then compute $\tilde{E}$ s.t. $|\tilde{E}-E|<\frac{b}{2}$.
- If $E=\uparrow$ then $\tilde{E}=\uparrow$.
- If $|\tilde{E}| \geq \frac{b}{2}$ then $\operatorname{sign}(E)=\operatorname{sign}(\tilde{E})$ else $E=0$.


## C. Yap's approach (exact geometric computation)

Approximate expression evaluation

- Given $E$ and a precision $p \in \mathbb{R}_{+}$compute an approximation of $E$ within precision $p$ !
- all $E$ are "programs", rooted, labeled directed acyclic graphs (DAG).
- use precision-driven approach.
- propagate precision values down to the leaves.
- approximate the value at the leaf to any desired precision.
- propate the approximations up to the root.

Numerical filters

- Numerical filters are an effective technique for speeding up predicate evaluation.


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## Shortly about our approach

- Let $I$ the set of coefficient vectors of polynomials of fixed degree, $O$ a discrete set.
- Let $E: I \rightarrow O$ the symbolic algorithm s.t.

Given $f \in I$ assigns $E(f)$ the invariants of the curve defined by $f$. This problem is ill-posed!

## Shortly about our approach

- Let $I$ the set of coefficient vectors of polynomials of fixed degree, $O$ a discrete set.
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Given $f \in I$ assigns $E(f)$ the invariants of the curve defined by $f$.
This problem is ill-posed!

- Let $A: I \times \mathbb{R}_{+} \rightarrow O$ the symbolic-numeric algorithm we designed s.t.

Given $(f, \epsilon) \in I \times \mathbb{R}_{+}$assigns the $\delta$-invariants of the curve defined by $f$.

- For $f \in I$ a perturbation of $f$ is a function
$f_{-}: \mathbb{R}_{+} \rightarrow I, \delta \mapsto f_{\delta}$ such that $\left|f-f_{\delta}\right| \leq \delta$ for all $\delta \in \mathbb{R}_{+}$.
We call $f$ the exact data, $f_{\delta}$ the perturbed data, $\delta$ the noise level (error, tolerance).
In our problem, we are given $f_{\delta}$ and $\delta$ but not $f$ !


## Shortly about our approach

We can prove the following properties of $A$ :

- $A_{-}$depends continuously on $f_{\delta}$ continuity (1).
- $\exists \alpha: \mathbb{R}_{+} \rightarrow \mathbb{R}_{+}$continuous, monotonic, $\lim _{\delta \rightarrow 0} \alpha(\delta)=0$ s.t. for any $f_{\delta}$

$$
\lim _{\delta \rightarrow 0} A\left(f_{\delta}, \alpha(\delta)\right)=E(f) \text {, i.e. convergence for perturbed data (2). }
$$

- In this case $\alpha$ is called the "parameter choice rule"!

The algorithm $A_{\epsilon}$ is called a regularization.
Instead of looking for the exact solution, we look for approximations with (1), (2).


Thank you for your attention.


Doctoral Program

