



Structured condition numbers for multiple eigenvalues

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Joint work with

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Structured Numerical Linear Algebra Problems:
Analysis, Algorithms and Applications

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Consider the **complex skew-symmetric** matrix

$$A = \begin{bmatrix} 0 & 1 & 0 & 1 \\ -1 & 0 & -i & 0 \\ 0 & i & 0 & i \\ -1 & 0 & -i & 0 \end{bmatrix},$$

which has one single eigenvalue $\lambda_0 = 0$ and Jordan form

$$J_3(0) \oplus J_1(0).$$



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- If A is subject to a **small** perturbation

$$A(\varepsilon) = A + \varepsilon E, \quad \varepsilon \ll 1,$$

with E an **arbitrary** 4×4 complex matrix, then $A(\varepsilon)$ has **generically** three eigenvalues of order $O(\varepsilon^{1/3})$, and one of order $O(\varepsilon)$.



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An example to begin with

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Sensitivity under **structured** perturbation **qualitatively** different than under **unstructured** perturbation



- 1) Hölder condition numbers (structured & unstructured) for multiple eigenvalues
- 2) Comparing structured and unstructured condition numbers for
 - 2.1) **generic** structured perturbations
 - 2.2) **nongeneric** structured perturbations
- 3) Concluding remarks



- Let λ_0 be a **simple** eigenvalue of $A \in \mathbb{C}^{n \times n}$, $\|\cdot\|$ matrix 2-norm.

Definition

The **condition number** of λ_0 is

$$\kappa(\lambda_0) = \lim_{\varepsilon \rightarrow 0} \sup \left\{ \frac{|\Delta\lambda|}{\varepsilon} : E \in \mathbb{C}^{n \times n}, \|E\| \leq 1, \lambda_0 + \Delta\lambda \in \text{sp}(A + \varepsilon E) \right\}$$



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- If x (resp. y) right (resp. left) e-vector corresp. to λ_0 with $y^H x = 1$, then

$$\kappa(\lambda_0) = \|x\| \|y\|$$



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- If x (resp. y) right (resp. left) e-vector corresp. to λ_0 with $y^H x = 1$, then

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But: if λ_0 is defective, then generically

$$\frac{\Delta\lambda}{\varepsilon} \rightarrow \infty \quad \text{as } \varepsilon \rightarrow 0 \quad \implies \quad \text{Need } \neq \text{definition for } \kappa(\lambda_0)$$



If the eigenproblem has some special **structure** (symmetric, skew-symmetric, Toeplitz, Hankel, zero patterns, symplectic, Hamiltonian,...)

- look for numerical algorithms that **preserve the structure** and spectral properties of the problem
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Hence, if $A \in \mathbb{S}$, class of **structured matrices**, \implies measure sensitivity **with respect to perturbations** $E \in \mathbb{S}$.

Leads to structured condition numbers.

$$\kappa(\lambda_0, \mathbb{S}) = \lim_{\varepsilon \rightarrow 0} \sup \left\{ \frac{|\Delta\lambda|}{\varepsilon} : E \in \mathbb{S}, \|E\| \leq 1, \lambda_0 + \Delta\lambda \in \text{sp}(A + \varepsilon E) \right\}.$$



Structured perturbations

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Question:

Is $\kappa(\lambda_0, \mathbb{S})$ much smaller than $\kappa(\lambda_0)$?



Many relevant contributions on **structured condition numbers** of **simple** eigenvalues

[Tisseur '03]

[Byers & Kressner '03]

[Noschese & Pasquini '06, '07]

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How about condition numbers for **multiple, defective** eigenvalues?



- Let λ_0 be a **multiple** e-value of $A \in \mathbb{C}^{n \times n}$, and let $n_1 \equiv$ size of largest Jordan block corresp. to λ_0 .

Then, the **worst-case** behaviour of λ_0 under small perturbations $A + \varepsilon E$ corresponds to

$$\hat{\lambda}(\varepsilon) = \lambda_0 + (\xi_k)^{1/n_1} \varepsilon^{1/n_1} + o(\varepsilon^{1/n_1}),$$

where ξ_k are the **eigenvalues** of a product $Y^H E X$

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- Let

$$P^{-1}AP = \begin{bmatrix} J_0 & 0 \\ 0 & * \end{bmatrix}$$

be a Jordan form of A , where J_0 gathers all r_1 Jordan blocks of size n_1 corresp. to λ_0 . Then

$$Y^H E X \in \mathbb{C}^{r_1 \times r_1}, \quad \text{where}$$

- X contains those **columns of P** which are **right** e-vectors of A corresp. to **blocks of largest size n_1** in J_0 .
- Y^H contains those **rows of P^{-1}** which are **left** e-vectors of A corresp. to **blocks of largest size n_1** in J_0 .



Lidskii's theory shows that **worst-case** behavior is

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Definition [M., Burke & Overton '97]

λ_0 **multiple** e-value of $A \in \mathbb{C}^{n \times n}$,

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$Y \equiv$ **left e-vectors** taken from all $n_1 \times n_1$ Jordan blocks J_{λ_0} .

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The **Hölder condition number** of λ_0 is the pair $\kappa(\lambda_0) = (n_1, \alpha)$, where

$$\alpha = \sup_{\substack{E \in \mathbb{C}^{n \times n} \\ \|E\| \leq 1}} \rho(Y^H E X), \quad \rho(\cdot) \equiv \text{spectral radius}$$



Hölder condition numbers for multiple eigenvalues

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One can prove that **for any unitarily invariant** matrix norm,

$$\alpha = \|XY^H\|_2$$



Again, if $\lambda_0 \in \text{sp}(A)$ for $A \in \mathbb{S}$, class of **structured** matrices, define **structured** Hölder condition number as a pair

$$\kappa(\lambda_0, \mathbb{S}) = (n_{\mathbb{S}}, \alpha_{\mathbb{S}}),$$

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Our goal: Determine and **compare** the **structured** and **unstructured** condition numbers of **defective** e-values for particular classes \mathbb{S} of matrices, e.g.,

complex symmetric, skew-symmetric, persymmetric, skew-Hermitian, Toeplitz, symmetric Toeplitz, Hankel, zero-structured, Hamiltonian, skew-Hamiltonian, symplectic,...



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Important:

$n_{\mathbb{S}}$ may be **strictly smaller** than n_1 ,
e.g. as in the initial 4×4 complex skew-symmetric example,
where $n_1 = 3$, $n_{\mathbb{S}} = 2$, and
 $\mathbb{S} \equiv$ skew-symmetric matrices



$$\kappa(\lambda_0) = (n_1, \alpha), \quad \kappa(\lambda_0, \mathbb{S}) = (n_{\mathbb{S}}, \alpha_{\mathbb{S}}).$$

First, consider the **generic** situation when

$$n_{\mathbb{S}} = n_1,$$

i.e., there is some $E \in \mathbb{S}$ giving rise to a perturbation expansion of order $O(\varepsilon^{1/n_1})$.



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$$\alpha = \sup_{\substack{E \in \mathbb{C}^{n \times n} \\ \|E\|_2 \leq 1}} \rho(Y^H E X) = \|XY^H\|_2, \quad \alpha_{\mathbb{S}} = \sup_{\substack{E \in \mathbb{S} \\ \|E\|_2 \leq 1}} \rho(Y^H E X)$$

and we want to know whether $\alpha_{\mathbb{S}} \ll \alpha$ or not.



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and we want to know whether $\alpha_{\mathbb{S}} \ll \alpha$ or not.

Usually, look first for some $E_{\mathbb{S}} \in \mathbb{S}$ such that

$$\rho(Y^H E_{\mathbb{S}} X) \approx \|XY^H\|.$$

Then,

$$\kappa(\lambda_0, \mathbb{S}) \approx \kappa(\lambda_0).$$



Structured Jordan form for $A \in \mathbb{S}$.

(see [Thompson'91], [Mehl'06], ...)



Induced structure in XY^H .



Mapping theorems: $E_0 u = \beta v$ with $|\beta| = 1$, $\|E_0\|_2 \approx 1$, $E_0 \in \mathbb{S}$.

(see [Rump'03], [Mackey,Mackey&Tisseur'06], ...)



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Take, for instance, $\mathbb{S} \equiv$ complex symm. matrices. Then

$$Y = \bar{X}, \quad \text{i.e.,} \quad \alpha = \|XY^H\|_2 = \|XX^T\|_2.$$

Let $XX^T = U\Sigma U^T$ be a Takagi factorization (i.e., an SVD), and let $u_1 = Ue_1$. Set $E_0 = \bar{u}_1 u_1^H \in \mathbb{S}$. Then,

$$\rho(E_0 XX^T) = \rho(\bar{u}_1 u_1^H XX^T) = \rho(u_1^H XX^T \bar{u}_1) = \sigma_{\max}(XX^T) = \|XX^T\|_2 = \alpha$$

Mapping: $E_0 u_1 = \bar{u}_1 u_1^H u_1 = \bar{u}_1$.



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Works well for *complex* Toeplitz, Hankel, persymmetric, Hermitian, symmetric, *real* Hamiltonian, skew-Hamiltonian,...

Does **not** work for *complex skew-symmetric* \longrightarrow can still be done using 'ad hoc' techniques



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For more details (including matrix pencils & matrix polynomials) see

D. Kressner, M. J. Peláez, and J. Moro. Structured Hölder condition numbers for multiple eigenvalues, preprint, 2006.



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What about **nongeneric** perturbations,
like *skew-symmetric* ones in the initial example?



Def: Given λ_0 e-value of A , and Y, X matrices of **left** and **right** e-vectors as before, a class \mathbb{S} of structured matrices is **nongeneric** if $n_{\mathbb{S}} < n_1$ or, equivalently, if

$$\sup_{\substack{E \in \mathbb{S} \\ \|E\|_2 \leq 1}} \rho(Y^H E X) = 0$$



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For instance, in our initial 4 by 4 example, both $Y = y$ and $X = x$ are vectors, since there is one single largest Jordan block of size 3.

Moreover, $y = \bar{x}$ since A is complex skew-symmetric, so

$$Y^H E X = y^H E x = x^T E x = 0 \quad \text{for any skew-symmetric } E \in \mathbb{S}.$$



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Given λ_0 e-value of A , and Y, X matrices of left and right e-vectors as before, a class \mathbb{S} of structured matrices is said to be **fully nongeneric** if

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Nongeneric and fully nongeneric structures

Def: Given λ_0 e-value of A , and Y, X matrices of left and right e-vectors as before, a class \mathbb{S} of structured matrices is **nongeneric** if $n_{\mathbb{S}} < n_1$ or, equivalently, if

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Can we somehow characterize **fully nongeneric** structures?



Let \mathbb{S} be a **linear structure**, i.e. \mathbb{S} is a linear subspace of $\mathbb{C}^{n \times n}$.

Def: Let \mathbb{S} be a **linear subspace** of $\mathbb{C}^{n \times n}$. Then, the **skew-structure** associated with \mathbb{S} is defined as

$$\text{Skew}(\mathbb{S}) = \{B \in \mathbb{C}^{n \times n} : \text{vec}(B)^H \text{vec}(A) = 0 \quad \forall A \in \mathbb{S}\},$$

where $\text{vec} \equiv$ stacking operator.

Theorem [Peláez & M. '08]

Let λ_0 be an e-value of A with e-vector matrices X and Y . Let y_i, x_j be, respectively, the columns of Y and X , and let \mathbb{S} be a linear structure. Then \mathbb{S} is fully nongeneric for λ_0 if and only if

$$y_i^H x_j \in \text{Skew}(\mathbb{S}) \quad \text{for every } i, j$$

The *Skew* operator produces all the ‘customary’ skew-families:



Some linear structures and their skew-structures

\mathbb{S}	$Skew(\mathbb{S})$
Symmetric	Skewsymmetric
Pseudo Symmetric	Pseudo Skewsymmetric
Persymmetric	Perskewsymmetric
Hamiltonian	Skew-Hamiltonian
Hermitian	Skew-Hermitian
Pseudo-Hermitian	Pseudo-Skew-Hermitian
Toeplitz	zero d-sums
Hankel	zero ad-sums
Circulant	zero ed-sums
Cocirculant	zero ead-sums



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$$\kappa(\lambda_0) = (n_1, \alpha), \quad \kappa(\lambda_0, \mathbb{S}) = (n_{\mathbb{S}}, \alpha_{\mathbb{S}}) \quad \text{with } n_{\mathbb{S}} < n_1,$$

where

- n_1 is the size of the largest λ_0 -Jordan block



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- How can we find $n_{\mathbb{S}}$?



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where

- n_1 is the size of the largest λ_0 -Jordan block
- How can we find $n_{\mathbb{S}}$?

Use the **Newton diagram**, a **geometric construction** which gives both the **leading exponents** and the **leading coefficients** in the asymptotic expansions of the *roots of a polynomial in two variables*.



For fully nongeneric structures,

$$\kappa(\lambda_0) = (n_1, \alpha), \quad \kappa(\lambda_0, \mathbb{S}) = (n_{\mathbb{S}}, \alpha_{\mathbb{S}}) \quad \text{with } n_{\mathbb{S}} < n_1,$$

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- How can we find $n_{\mathbb{S}}$?

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How does the Newton diagram work?

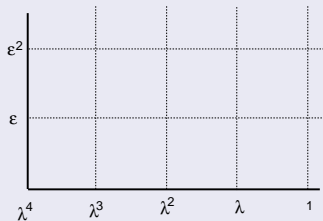


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$$p(\lambda, \varepsilon) = \lambda^4 + (2\varepsilon - \varepsilon^2)\lambda^3 + \varepsilon^2\lambda^2 + (\varepsilon - \varepsilon^3)\lambda + \varepsilon^2$$

Draw a cartesian grid and label the axes with λ, ε



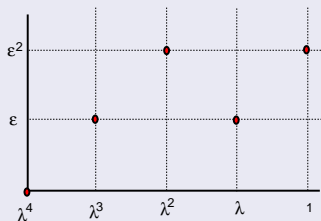
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Consider only the **dominant** terms

Step 1: plot a point for each dominant $\varepsilon^p \lambda^q$ terms



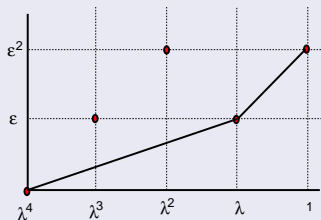
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Step 2: draw the lower boundary of the convex hull: that's the ND



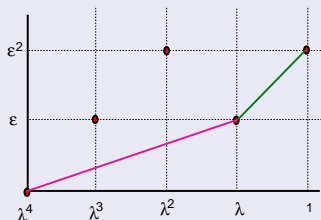
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Consider only the **dominant** terms

Step 1: each slope is a leading power in the Puiseux ε -expansion



three e-vals of $O(\varepsilon^{1/3})$, one e-val of $O(\varepsilon)$



The 4×4 example: leading exponents via the ND

Unstructured perturbation

$$\begin{bmatrix} 0 & 1 & 0 & 1 \\ -1 & 0 & -i & 0 \\ 0 & i & 0 & i \\ -1 & 0 & -i & 0 \end{bmatrix} + \varepsilon E, \quad E = \begin{bmatrix} e_{11} & e_{12} & e_{13} & e_{14} \\ e_{21} & e_{22} & e_{23} & e_{24} \\ e_{31} & e_{32} & e_{33} & e_{34} \\ e_{41} & e_{42} & e_{43} & e_{44} \end{bmatrix} \in \mathbb{C}^{4 \times 4}$$

If $P^{-1}AP = J$, then $P^{-1}(A + \varepsilon E)P = J + \tilde{E}$, where $J = J_3(0) \oplus J_1(0)$ and

$$\tilde{E} = P^{-1}EP = \left[\begin{array}{ccc|c} * & * & * & * \\ * & * & * & * \\ \hline \Phi & * & * & * \\ * & * & * & * \end{array} \right], \quad \text{with } \Phi = y^H E x,$$



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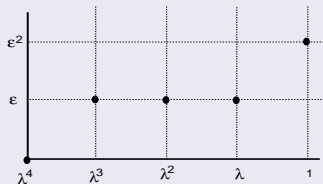
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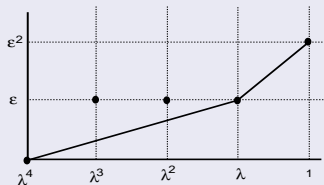
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Step 2: draw lower boundary of convex hull



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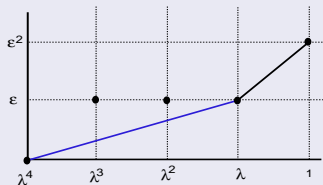
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Smallest slope $1/3 \Rightarrow n_1 = 3$



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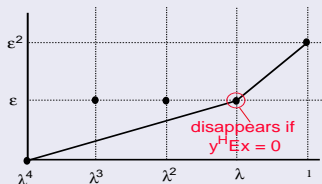
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The term in $\lambda \varepsilon$ is present only if $\Phi = y^H E x \neq 0$



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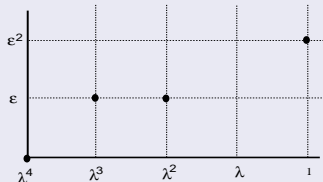
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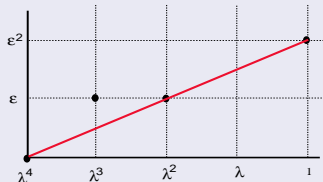
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Step 2: draw lower boundary of convex hull



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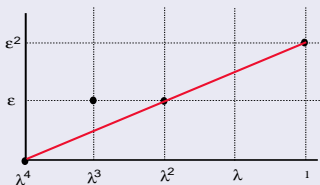
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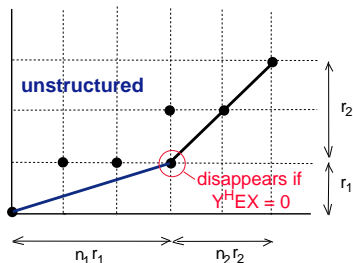
Step 2: draw lower boundary of convex hull



Only slope $1/2 \Rightarrow n_1 = 2$



Same can be done in general for arbitrary **fully nongeneric** structure:

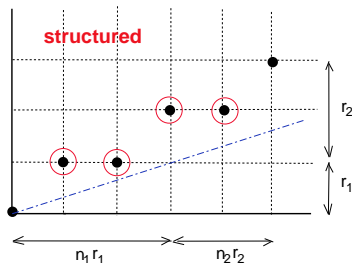


Bending point disappears from ND as soon as perturbations become fully nongeneric, i.e., when

$$Y^H EX = 0 \text{ for all } E \in \mathbb{S}.$$



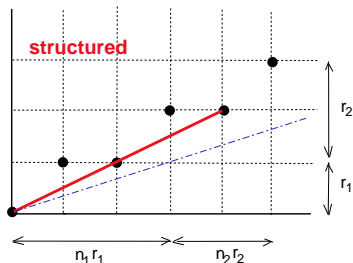
Same can be done in general for arbitrary **fully nongeneric** structure:



- For perturbations in \mathbb{S} , identify **most likely points** to lie on lowest segment in the ND
- Depend on sizes n_1, n_2 and numbers r_1, r_2 of λ_0 -Jordan blocks. Possibly, also on the structure \mathbb{S}



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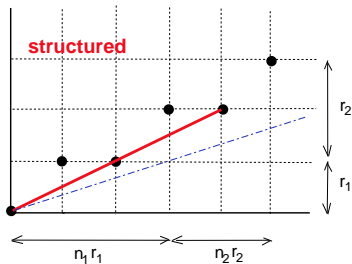
Finally, determine the reciprocal

$$n_S < n_1$$

of the lowest possible slope



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In this way, **explicit formulas** can be found for n_S , depending on the quantities n_1, n_2, r_1, r_2

- **Entry-wise information** on the structure is important, to assess which points are present on the grid when the perturbations are structured.
- Otherwise, the formulas give just **upper bounds** on n_S



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- Still several relevant classes to be explored (e.g., zero structures).

