Query languages with structural and analytic properties

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1 MATLANG syntax and semantics

We assume that we have a supply of matrix variables. The definition of an instance I on MATLANG is a function defined on a nonempty set $var(I) = \{A, B, M, C, \ldots\}$, that assigns a concrete matrix to each element (matrix name) of var(I).

Every expression e is a matrix, either a matrix of var(I) (base matrix, if you will) or a result of an operation over matrices.

The syntax of MATLANG expressions is defined by the following grammar. Every sentence is an expression itself.

$$e = M \quad \text{(matrix variable)}$$

$$\text{let } M = e_1 \text{ in } e_2 \quad \text{(local binding)}$$

$$e^* \quad \text{(conjugate transpose)}$$

$$\mathbf{1}(e) \quad \text{(one-vector)}$$

$$\text{diag}(e) \quad \text{(diagonalization of a vector)}$$

$$e_1 \cdot e_2 \quad \text{(matrix multiplication)}$$

$$\text{apply } [f] (e_1, \dots, e_n) \quad \text{(pointwise application of } f)$$

The operations used in the semantics of the language are defined over complex numbers.

- Transpose: if A is a matrix then A^* is its conjugate transpose.
- One-vector: if A is a $n \times m$ matrix then $\mathbf{1}(A)$ is the $m \times 1$ column vector full of ones.
- **Diag:** if v is a $m \times 1$ column vector then diag(v) is the matrix

$$\begin{bmatrix} v_1 & 0 & 0 & \dots & 0 \\ 0 & v_2 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & v_m \end{bmatrix}$$

- Matrix multiplication: if A is a $n \times m$ matrix and B is a $m \times p$ matrix then $A \cdot B$ is the $n \times p$ matrix with $(AB)_{ij} = \sum_{k=1}^{n} A_{ik} B_{kj}$.
- Pointwise application: if $A^{(1)}, \ldots, A^{(n)}$ are $m \times p$ matrices, then apply $[f](A^{(1)}, \ldots, A^{(n)})$ is the $m \times p$ matrix C where $C_{ij} = f(A_{ij}^{(1)}, \ldots, A_{ij}^{(n)})$.

The formal semantics have a set of rules for an expression e to be valid on an instance I, this is, e successfully evaluates to a matrix A on the instance I. This success is denoted as e(I) = A. Here I[M := A] denotes the instance that is equal to I except that maps M to the matrix A.

$$\begin{array}{c|cccc} \mathbf{Expression} & \mathbf{Condition\ for\ validity} \\ & (\text{let}\ M = e_1\ \text{in}\ e_2)(I) = B & e_1(I) = A,\ e_2(I\ [M := A]) = B \\ & e^*(I) = A^* & e(I) = A \\ & \mathbf{1}(e)(I) = \mathbf{1}(A) & e(I) = A \\ & \text{diag}(e)(I) = \text{diag}(A) & e(I) = A,\ A \ \text{is a column vector} \\ & e_1 \cdot e_2(I) = A \cdot B & \# \ \text{columns of}\ A = \# \ \text{rows of}\ B \\ & \text{apply}[f]\ (e_1, \dots, e_n)(I) = \text{apply}\ [f]\ (A_1, \dots, A_n) & \forall k, e_k(I) = A \ \text{and all}\ A_k \ \text{have the same dimentions} \\ \end{array}$$

2 Adding canonical vectors to MATLANG

One thing that we cannot do in MATLANG is to obtain a specific entry of a matrix. This entry is expected to be a 1×1 matrix. We can do this by adding the standard unit vectors e_j where

$$e_i = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \rightarrow i\text{-th position}$$

We know show some examples of what can we express with this new feature. For ilustrative reasons, we asume that all the dimentions are well suited for the corresponding operation.

- Get A_{ij} with $e_i^* \cdot A \cdot e_j$.
- The expression $e_i \cdot e_j^*$ is the matrix that has a 1 in the position i, j and zero everywhere else.
- Given a vector v, the expression $v \cdot e_i^*$ is the matrix

$$\begin{bmatrix} 0 & \cdots & v_1 & \cdots & 0 \\ 0 & \cdots & v_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & v_n & \cdots & 0 \end{bmatrix}$$

- Replace column j of A with zeros: $A(I e_j \cdot e_j^*)$.
- Replace column j of A with a vector v: $A(I e_j \cdot e_j^*) + v \cdot e_j^*$.

Note that $I = \text{diag}(\mathbf{1}(A))$ and the sum of matrices can be implemented as apply [+](A, B).

3 Connection with logic

4 Core of Matlab and R

MATLAB

The basic operations of MATLAB over matrices are:

• mldivide(A, B): returns x such that Ax = B.

- descomposition(A): returns a decomposition or factorization LU, LDL, QR, Cholesky, etc.
- inv(A): returns A^{-1} .
- multiplication: compute $A \cdot B$.
- transpose(A): returns A^T .
- conjugate transpose(A): returns A'.
- matrix power(A, k): returns A^k .
- eigen(A): returns the eigenvectors and the eigenvectors matrices of A.
- funm(A, f): returns matrix B with elements $b_{ij} = f(a_{ij})$.
- **crossprod**(a, b): vectorial product, returns c such that $c \perp a, b$.
- **dotprod**(a,b): returns $a \cdot b$.
- $\operatorname{diag}(v)$: v vector. Returns the following matrix:

$$\begin{bmatrix} v_1 & 0 & 0 & \dots & 0 \\ 0 & v_2 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & v_n \end{bmatrix}$$

• diag(A): given matrix A, it returns

$$\begin{bmatrix} a_{11} \\ a_{22} \\ \vdots \\ a_{nn} \end{bmatrix}$$

- $\det(A)$: returns the determinant of A.
- **zeros**(n, m): returns a $n \times m$ matrix full of zeros.
- ones(n, m): returns a $n \times m$ matrix full of ones.
- A[i, j]: you can get A_{ij} .

\mathbf{R}

The basic operations of the language R over matrices are:

- A%*%B: matrix multiplication.
- **A*B**: pointwise multiplication.
- t(A): transpose.
- diag(v): returns the matrix

$$\begin{bmatrix} v_1 & 0 & 0 & \dots & 0 \\ 0 & v_2 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & v_n \end{bmatrix}$$

• diag(A): Returns the vector

$$\begin{bmatrix} a_{11} \\ a_{22} \\ \vdots \\ a_{nn} \end{bmatrix}$$

- $\operatorname{diag}(\mathbf{k})$: k scalar. It creates the $k \times k$ identity matrix.
- matrix(k, n, m): returns the $n \times m$ matrix, where every entry is equal to k.
- solve(A, b): returns x such that Ax = b.
- solve(A): returns A^{-1} .
- det(A): determinant of A.
- y < -eigen(A): stores de eigenvalues of A in y\$val and the eigenvectors in y\$vec.
- y < -svd(A): it computes and stores the following:
 - y\$d: vector of the singular values of A.
 - y\$u: matrix of the left singular vectors of A.
 - y\$v: matrix of the right singular vectors of A.
- $\mathbf{R} < -\mathbf{chol}(\mathbf{A})$: Cholesky fatorization, R'R = A.
- $\mathbf{y} < -\mathbf{qr}(\mathbf{A})$: QR decomposition, strong in y\$qr.
- cbind(A,B, v, ...): joins matrices and vector horizontally, returns a matrix.
- rbind(A,B, v, ...): joins matrices and vector vertically, returns a matrix.
- rowMeans(A): returns the vector of the averages over the rows of A.
- colMeans(A): returns the vector of the averages over the columns of A.
- rowSums(A): returns the vector of the sums over the rows of A.
- colSums(A): returns the vector of the sums over the columns of A.
- outer(A, B, f): applies $f(\cdot, \cdot)$. Returns matrix C of entries $c_{ij} = f(a_{ij}, b_{ij})$.
- A[i, j]: you can get A_{ij} .