

4.5 Symbol Table Applications

Set ADT: unordered collection of distinct keys.

- **Insert** a key.
- **Check** if set contains a given key.
- **Delete** a key.

SET interface.

- `add(key)` insert the key
- `contains(key)` is given key present?
- `remove(key)` remove the key
- `iterator()` return iterator over all keys

Q. How to implement?

Java library: `java.util.HashSet`.

Set Client: Remove Duplicates

Remove duplicates. [e.g., from commercial mailing list]

- Read in a key.
- If key is not in set, insert and print it out.

```
public class DeDup {
    public static void main(String[] args) {
        SET<String> set = new SET<String>();
        while (!StdIn.isEmpty()) {
            String key = StdIn.readString();
            if (!set.contains(key)) {
                set.add(key);
                System.out.println(key);
            }
        }
    }
}
```

More Set Applications

Application	Purpose	Key
Spell checker	Identify misspelled words	Word
Browser	Highlight previously visited pages	URL
Chess	Detect repetition draw	Board position
Spam blacklist	Prevent spam	IP addresses of spammers

Extensions.

- Ignore stopwords: the, on, of, etc.
- Multi-word queries:
 - set intersection (AND)
 - set union (OR).
- Record position and number of occurrences of word in document.

Sparse Vectors and Matrices

Vectors and Matrices

Vector. Ordered sequence of N real numbers.

Matrix. N-by-N table of real numbers.

$$\begin{aligned}
 a &= [0 \ 3 \ 15], \quad b = [-1 \ 2 \ 2] \\
 a + b &= [-1 \ 5 \ 17] \\
 a \circ b &= (0 \cdot -1) + (3 \cdot 2) + (15 \cdot 2) = 36
 \end{aligned}$$

$$A = \begin{bmatrix} 0 & 1 & 1 \\ 2 & 4 & -2 \\ 0 & 3 & 15 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}, \quad A + B = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 6 & -2 \\ 0 & 3 & 18 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 & 1 \\ 2 & 4 & -2 \\ 0 & 3 & 15 \end{bmatrix} \times \begin{bmatrix} -1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \\ 36 \end{bmatrix}$$

Sparsity

Def. An N-by-N matrix is **sparse** if it has $O(N)$ nonzeros entries.

Empirical fact. Large matrices that arise in practice are usually sparse.

Matrix representations.

- 2D array: space proportional to N^2 .
- Goal: space proportional to number of nonzeros, without sacrificing fast access to individual elements.

Ex. Google performs matrix-vector product with $N = 4$ billion!

Sparse Vector Implementation

```
public class SparseVector {
    private final int N;
    private ST<Integer, Double> st = new ST<Integer, Double>();

    public SparseVector(int N) { this.N = N; }

    public void put(int i, double value) {
        if (value == 0.0) st.remove(i);
        else st.put(i, value);
    }

    public double get(int i) {
        if (st.contains(i)) return st.get(i);
        else return 0.0;
    }
}
```

$a[i] = \text{value}$

$a[i]$

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Sparse Vector Implementation (cont)

```
// return a · b
public double dot(SparseVector b) {
    SparseVector a = this;
    double sum = 0.0;
    for (int i : a.st)
        if (b.st.contains(i)) sum += a.get(i) * b.get(i);
    return sum;
}
```

```
// return c = a + b
public SparseVector plus(SparseVector b) {
    SparseVector a = this;
    SparseVector c = new SparseVector(N);
    for (int i : a.st) c.put(i, a.get(i));
    for (int i : b.st) c.put(i, b.get(i) + c.get(i));
    return c;
}
```

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Sparse Matrix Implementation

```
public class SparseMatrix {
    private final int N;
    private SparseVector[] rows;

    public SparseMatrix(int N) {
        this.N = N;
        rows = new SparseVector[N];
        for (int i = 0; i < N; i++)
            rows[i] = new SparseVector(N);
    }

    public void put(int i, int j, double value) {
        rows[i].put(j, value);
    }

    public double get(int i, int j) {
        return rows[i].get(j);
    }
}
```

N-by-N matrix
of all zeros

$A[i][j] = \text{value}$

$A[i][j]$

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Sparse Matrix Implementation (cont)

```
// return b = A*x
public SparseVector times(SparseVector x) {
    SparseMatrix A = this;
    SparseVector b = new SparseVector(N);
    for (int i = 0; i < N; i++)
        b.put(i, rows[i].dot(x));
    return b;
}
```

```
// return C = A + B
public SparseMatrix plus(SparseMatrix B) {
    SparseMatrix A = this;
    SparseMatrix C = new SparseMatrix(N);
    for (int i = 0; i < N; i++)
        C.rows[i] = A.rows[i].plus(B.rows[i]);
    return C;
}
```

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A Plan for Spam

Bayesian spam filter.

- Filter based on analysis of previous messages.
- User trains the filter by classifying messages as spam or ham.
- Parse messages into tokens (alphanumeric, dashes, ', \$)

Build data structures.

- Hash table A of tokens and frequencies for spam.
- Hash table B of tokens and frequencies for ham.
- Hash table C of tokens with probability p that they appear in spam.

```
double h = 2.0 * ham.freq(word);
double s = 1.0 * spam.freq(word);
double p = (s/spams) / (h/hams + s/spams);
```

bias probabilities to avoid false positives

Reference: <http://www.paulgraham.com/spam.html>

A Plan for Spam

Identify incoming email as spam or ham.

- Find 15 most interesting tokens (difference from 0.5).
- Combine probabilities using Bayes law. which data structure?

$$\frac{p_1 \times p_2 \times \dots \times p_{15}}{(p_1 \times p_2 \times \dots \times p_{15}) + ((1 - p_1) \times (1 - p_2) \times \dots \times (1 - p_{15}))}$$

- Declare as spam if threshold > 0.9.

Details.

- Words you've never seen.
- Words that appear in ham corpus but not spam corpus, vice versa.
- Words that appear less than 5 times in spam and ham corpuses.
- Update data structures.