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#### Efficient Calculation of Second Degree Polynomial Features on Sparse Matrices

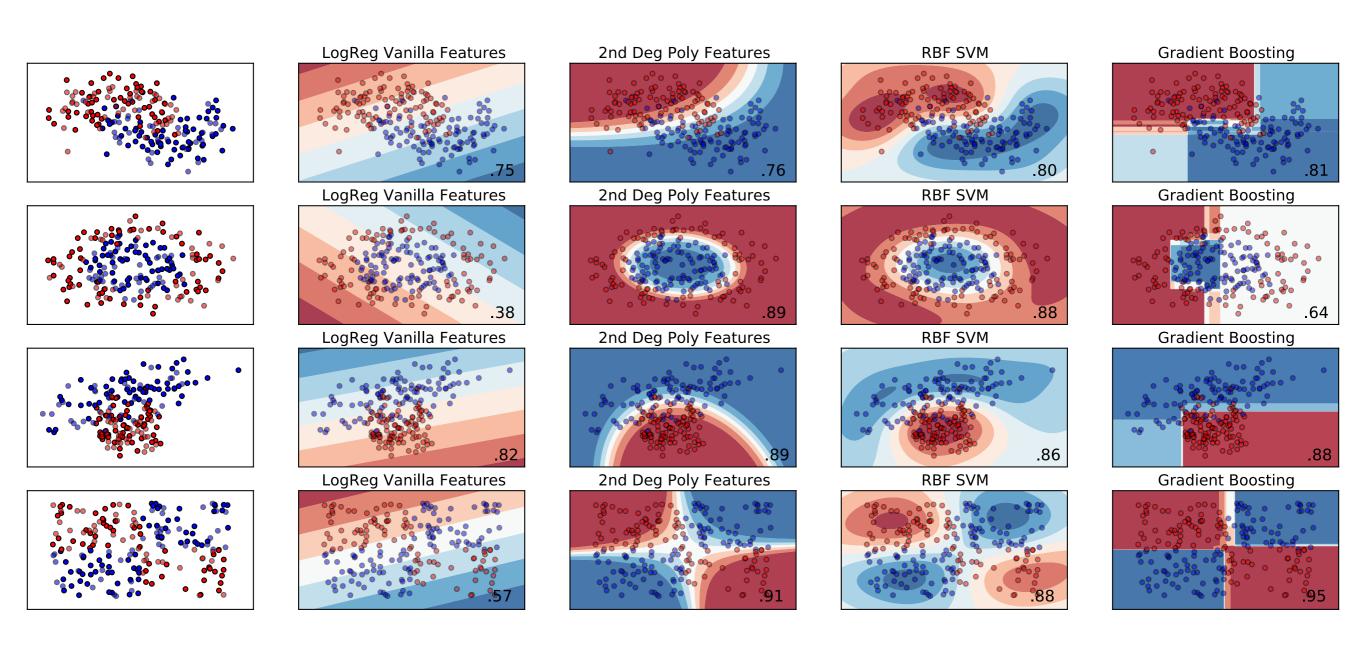
## Handling Nonlinearity

- \* Work in kernel space, e.g. rbf kernel
- \* Nonlinear model, e.g. gradient boosting
- \* Apply an input transformation
  - \* e.g.  $x_i \rightarrow \cos(x_i)$  for all components

## Polynomial Features

- \* Products of combinations of features
  - \*  $\{x_i \cdot x_j : i, j \in \{0, 1, ..., D-1\} \land i \leq j\}$
- \* [a, b, c] has polynomial features [a<sup>2</sup>, ab, ac, b<sup>2</sup>, bc, c<sup>2</sup>]
- \* D features, (D<sup>2</sup>+D)/2 2nd degree polynomial features

# What they can do (graphically)



#### Calculation

- \* Naive Calculation
  - \* Walk through all pairs, take products
  - Zero whenever either inputs are zero
  - \* Unnecessary calculations when matrix is sparse
- \* Smarter Calculation
  - Take products of non-zero elements
  - \* Problem: knowing output column index given input indices

## Column Output Location

```
\begin{aligned} &\{x_i \cdot x_j : i, j \in \{0, 1, ..., D-1\} \land i \leq j\} \\ &\{(i, j) : i, j \in \{0, 1, ..., D-1\} \land i \leq j\} \rightarrow \{0, 1, ..., \frac{D^2 + D}{2} - 1\} \end{aligned}
```

$$\vec{x}$$
 $\vec{x}$ 
 $x_0$ 
 $x_1$ 
 $x_2$ 
 $x_3$ 
 $x_4$ 
 $x_4$ 
 $x_5$ 
 $x_6$ 
 $x_1$ 
 $x_2$ 
 $x_3$ 
 $x_4$ 
 $x_5$ 
 $x_5$ 
 $x_6$ 
 $x_7$ 
 $x_8$ 
 $x_8$ 
 $x_9$ 
 $x_$ 

polynomial-index
$$(i, j|D) = \frac{2Di - i^2 + 2j - 3i - 2}{2} + i + 1$$

## Algorithm Comparison

```
Sparse Polynomials(A)
Dense Polynomials(A)
                                                     map(a,b) = \frac{2Da - a^2 + 2b - 3a - 2}{2} + a + 1
    N = \text{row count of } A
    D = \text{column count of } A
                                                     N = \text{row count of } A
                                                     D = \text{column count of } A
    B = \text{Matrix of size } N \times \frac{D^2 + D}{2}
                                                     B = \text{Compressed Sparse Row Matrix of size } N \times \frac{D^2 + D}{2}
    for row in A
                                                     for row in A
         k = 0
                                                           N_{zc} = nonzero columns of row
         for 1 = 0 to D
                                                           for i=0 to |N_{zc}|
               for j = i to D
                                                                 for j = i to |N_{zc}|
                     B[i,k] = row[i] \cdot row[j]
                                                                      k = map(i, j)
                     k = k + 1
                                                                      r = index of row
```

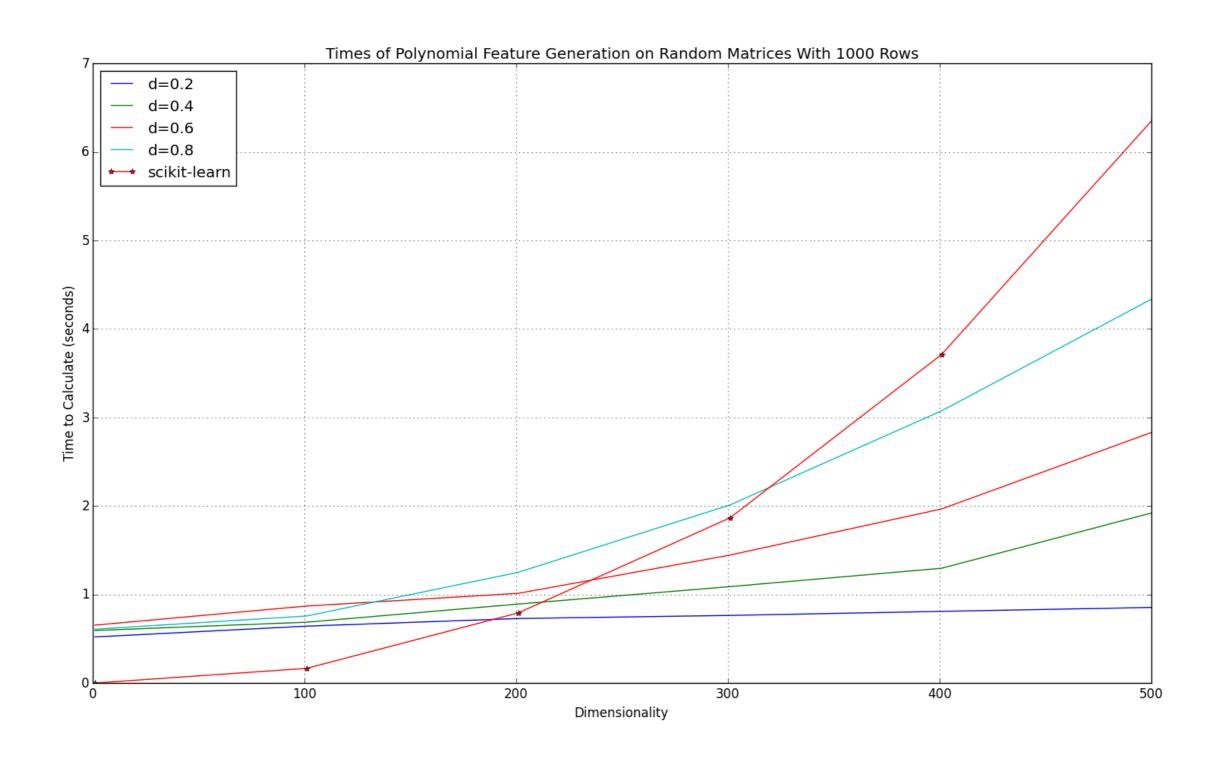
 $O(ND^2)$ ,  $\Theta(ND^2)$ 

 $O(ND^2)$ ,  $\Theta(Nd^2D^2)$  where d is density

 $B[r,k] = row[i] \cdot row[j]$ 

Average complexities decrease quadratically w.r.t. density!

#### How fast is it?



# Real Usage

dataset	instances	features	polynomial features	density	dense space	sparse space	dense time	sparse time
20 Newsgroups	11,314	130,107	8,463,980,778	0.12	scikit-learn: MemoryError	5333 MB	scikit-learn: MemoryError	109 s
Connect Four	67,557	126	8,127	0.11	4191 MB	735 MB	26 s	44 s

#### Code available on github.com/AWNystrom

