

CSEP 590
Data Compression
Autumn 2007

Predictive Coding
Burrows-Wheeler Transform

Predictive Coding

- The next symbol can be statistically predicted from the past.
 - Code with context
 - Code the difference
 - Move to front, then code
- Goal of prediction
 - The prediction should make the distribution of probabilities of the next symbol as skewed as possible
 - After prediction there is no way to predict more so we are in the first order entropy model

Bad and Good Prediction

- From information theory – The lower the information the fewer bits are needed to code the symbol.

$$\text{inf}(a) = \log_2\left(\frac{1}{P(a)}\right)$$

- Examples:
 - $P(a) = 1023/1024$, $\text{inf}(a) = .000977$
 - $P(a) = 1/2$, $\text{inf}(a) = 1$
 - $P(a) = 1/1024$, $\text{inf}(a) = 10$

Entropy

- Entropy is the expected number of bits to code a symbol in the model with a_i having probability $P(a_i)$.

$$H = \sum_{i=1}^m P(a_i) \log_2 \left(\frac{1}{P(a_i)} \right)$$

- Good coders should be close to this bound.
 - Arithmetic
 - Huffman
 - Golomb
 - Tunstall

PPM

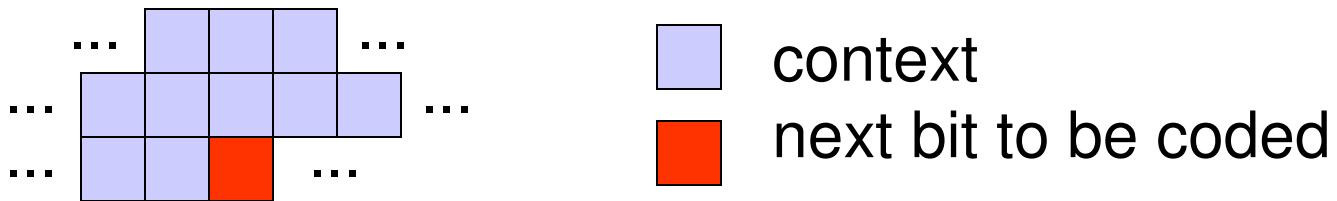
- Prediction with Partial Matching
 - Cleary and Witten (1984)
 - Tries to find a good context to code the next symbol

good	context	a	...	e	...	i	...	r	...	s	...	y
	the	0	0	5	7	4	7					
	he	10	1	7	10	9	7					
	e	12	2	10	15	10	10					
	<nil>	50	70	30	35	40	13					

- Uses adaptive arithmetic coding for each context

JBIG

- Coder for binary images
 - documents
 - graphics
- Codes in scan line order using context from the same and previous scan lines.



- Uses adaptive arithmetic coding with context

JBIG Example

	0	0	0	
0	0	0	0	0
0	0		0	

next bit	0	1
frequency	100	10

$$H = \frac{10}{110} \log\left(\frac{110}{10}\right) + \frac{100}{110} \log\left(\frac{110}{100}\right) = .44$$

	0	1	1	
0	1	1	1	0
0	1		0	

next bit	0	1
frequency	15	50

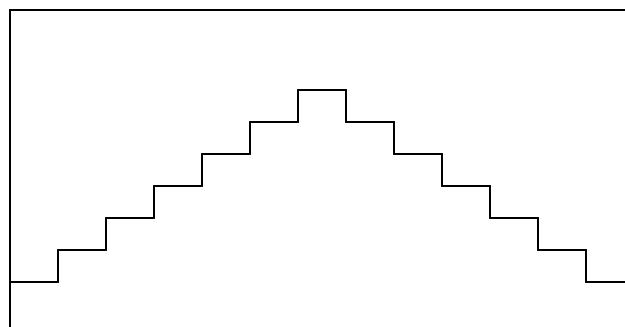
$$H = \frac{15}{65} \log\left(\frac{65}{15}\right) + \frac{50}{65} \log\left(\frac{65}{50}\right) = .78$$

Issues with Context

- Context dilution
 - If there are too many contexts then too few symbols are coded in each context, making them ineffective because of the zero-frequency problem.
- Context saturation
 - If there are too few contexts then the contexts might not be as good as having more contexts.
- Wrong context
 - Again poor predictors.

Prediction by Differencing

- Used for Numerical Data
- Example: 2 3 4 5 6 7 8 7 6 5 4 3 2



- Transform to 2 1 1 1 1 1 1 -1 -1 -1 -1 -1
– much lower first-order entropy

General Differencing

- Let x_1, x_2, \dots, x_n be some numerical data that is correlated, that is x_i is near x_{i+1}
- Better compression can result from coding
 $x_1, x_2 - x_1, x_3 - x_2, \dots, x_n - x_{n-1}$
- This idea is used in
 - signal coding
 - audio coding
 - video coding
- There are fancier prediction methods based on linear combinations of previous data, but these may require training.

Move to Front Coding

- Non-numerical data
- The data have a relatively small working set that changes over the sequence.
- Example: a b a b a a b c c b b c c c c b d b c c
- Move to Front algorithm
 - Symbols are kept in a list indexed 0 to m-1
 - To code a symbol output its index and move the symbol to the front of the list

Example

- Example: a b a b a a b c c b b c c c b d b c c
0

0	1	2	3
a	b	c	d

Example

- Example: a b a b a a b c c b b c c c b d b c c
0 1

0	1	2	3
a	b	c	d

↓

0	1	2	3
b	a	c	d

Example

- Example: a b a b a a b c c b b c c c b d b c c
0 1 1

0	1	2	3
b	a	c	d
↓			

0	1	2	3
a	b	c	d

Example

- Example: a b a b a a b c c b b c c c b d b c c
0 1 1 1

0	1	2	3
a	b	c	d
↓			

0	1	2	3
b	a	c	d

Example

- Example: a b a b a a b c c b b c c c b d b c c
0 1 1 1 1

0	1	2	3
b	a	c	d
↓			

0	1	2	3
a	b	c	d

Example

- Example: a b a b a a b c c c b b c c c c b d b c c
0 1 1 1 1 0

0	1	2	3
a	b	c	d

Example

- Example: a b a b a a b c c b b c c c c b d b c c
0 1 1 1 1 0 1

0	1	2	3
a	b	c	d

↓

0	1	2	3
b	a	c	d

Example

- Example: a b a b a a b c c b b c c c b d b c c
0 1 1 1 1 0 1 2

0	1	2	3
b	a	c	d
↓			

0	1	2	3
c	b	a	d

Example

- Example: a b a b a a b c c b b c c c c b d b c c
0 1 1 1 1 0 1 2 0 1 0 1 0 0 0 1 3 1 2 0

0	1	2	3
c	b	d	a

Example

- Example: a b a b a a b c c b b c c c c b d b c c
0 1 1 1 1 0 1 2 0 1 0 1 0 0 0 1 3 1 2 0

Frequencies of {a, b, c, d}

a	b	c	d
4	7	8	1

Frequencies of {0, 1, 2, 3}

0	1	2	3
8	9	2	1

Extreme Example

Input:

aaaaaaaaaaabbbbbbbbbbcccccccccddddd

Output

000000000010000000002000000000300000000

Frequencies of a b c d

a	b	c	d
10	10	10	10

Frequencies of 0 1 2 3

0	1	2	3
37	1	1	1

Burrows-Wheeler Transform

- Burrows-Wheeler, 1994
- BW Transform creates a representation of the data which has a small working set.
- The transformed data is compressed with move to front compression.
- The decoder is quite different from the encoder.
- The algorithm requires processing the entire string at once (it is not on-line).
- It is a remarkably good compression method.

Encoding Example

- abracadabra
- 1. Create all cyclic shifts of the string.

0	abracadabra
1	bracadabraaa
2	racadabraaab
3	acadabraabbr
4	cadabraabra
5	adabraabrac
6	dabraabrac
7	abraabracad
8	braabracada
9	raabracadab
10	aabracadabr

Encoding Example

2. Sort the strings alphabetically in to array A

0	abracadabra	A	0	aabracadabr
1	bracadabraa		1	abraabracad
2	racadabraab		2	abracadabra
3	acadabraabr		3	acadabraabbr
4	cadabraabra	→	4	adabraabrac
5	adabraabrac		5	braabracada
6	dabraabrac		6	bracadabraa
7	abraabracad		7	cadabraabra
8	braabracada		8	dabraabrac
9	raabracadab		9	raabracadab
10	aabracadabr		10	racadabraab

Encoding Example

3. $L =$ the last column

A		$L =$
0	aabracadabr	
1	abraabracad	
2	abracadabra	
3	acadabraabr	
4	adabraabrac	
5	braabracada	
6	bracadabraa	
7	cadabraabra	
8	dabraabrac	
9	raabracadab	
10	racadabraab	

Encoding Example

4. Transmit X the index of the input in A and L (using move to front coding).

A	
0	aabracadab r
1	abraabracad
2	abracadabra
3	acadabraab r
4	adabraab r c
5	braabracada
6	bracadabraa
7	cadabraabra
8	dabraabrac a
9	raabracadab
10	racadabraab

L = rdarcaaaabb
X = 2

Why BW Works

- Ignore decoding for the moment.
- The prefix of each shifted string is a context for the last symbol.
 - The last symbol appears just before the prefix in the original.
- By sorting, similar contexts are adjacent.
 - This means that the predicted last symbols are similar.

Decoding Example

- We first decode assuming some information. We then show how to compute the information.
- Let A^s be A shifted by 1

A	A^s
0 aabracadabr	0 raabracadab
1 abraabracad	1 dabraabrac
2 abracadabra	2 aabracadabr
3 acadabraab	3 racadabraab
4 adabraabrac	4 cadabraabra
5 braabracada	5 abraabracad
6 bracadabraa	6 abracadabra
7 cadabraabra	7 acadabraabr
8 dabraabrac	8 adabraabrac
9 raabracadab	9 braabracada
10 racadabraab	10 bracadabraa

Decoding Example

- Assume we know the mapping $T[i]$ is the index in A^s of the string i in A .
- $T = [2 \ 5 \ 6 \ 7 \ 8 \ 9 \ 10 \ 4 \ 1 \ 0 \ 3]$

A	A^s
0 aabracadabr	0 raabracadab
1 abraabracad	1 dabraabraca
2 abracadabra	2 aabracadabr
3 acadabraabbr	3 racadabraab
4 adabraabrac	4 cadabraabra
5 braabracada	5 abraabracad
6 bracadabraa	6 abracadabra
7 cadabraabra	7 acadabraabr
8 dabraabraca	8 adabraabrac
9 raabracadab	9 braabracada
10 racadabraab	10 bracadabraa

Decoding Example

- Let F be the first column of A , it is just L , sorted.

$$F = \begin{matrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ a & a & a & a & a & b & b & c & d & r & r \end{matrix}$$
$$T = \begin{matrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ 2 & 5 & 6 & 7 & 8 & 9 & 10 & 4 & 1 & 0 & 3 \end{matrix}$$

- Follow the pointers in T in F to recover the input starting with X .

Decoding Example

$$F = \begin{array}{cccccccccc} 0 & 1 & \underline{2} & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ a & a & a & a & a & b & b & c & d & r & r \end{array}$$

$$T = \begin{array}{cccccccccc} 0 & 1 & \underline{2} & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ 2 & 5 & 6 & 7 & 8 & 9 & 10 & 4 & 1 & 0 & 3 \end{array}$$

a

Decoding Example

$$F = \begin{array}{cccccccccc} 0 & 1 & \underline{2} & 3 & 4 & 5 & \underline{6} & 7 & 8 & 9 & 10 \\ a & a & a & a & a & b & b & c & d & r & r \end{array}$$

$$T = \begin{array}{cccccccccc} 0 & 1 & \underline{2} & 3 & 4 & 5 & \underline{6} & 7 & 8 & 9 & 10 \\ 2 & 5 & 6 & 7 & 8 & 9 & 10 & 4 & 1 & 0 & 3 \end{array}$$

ab

Decoding Example

$$F = \begin{array}{cccccccccc} 0 & 1 & \underline{2} & 3 & 4 & 5 & \underline{6} & 7 & 8 & 9 & \underline{10} \\ a & a & a & a & a & b & b & c & d & r & r \end{array}$$

$$T = \begin{array}{cccccccccc} 0 & 1 & \underline{2} & 3 & 4 & 5 & \underline{6} & 7 & 8 & 9 & \underline{10} \\ 2 & 5 & 6 & 7 & 8 & 9 & 10 & 4 & 1 & 0 & 3 \end{array}$$

abr

Decoding Example

- Why does this work?
- The first symbol of $A[T[i]]$ is the second symbol of $A[i]$ because $A^s[T[i]] = A[i]$.

A	T	A^s
0	aabracadabr	2
1	abraabracad	5
2	abracadabra	6
3	acadabraabr	7
4	adabraabrac	8
5	braabracada	9
6	bracadabraa	10
7	cadabraabra	4
8	dabraabraca	1
9	raabracadab	0
10	racadabraab	3

Decoding Example

- How do we compute F and T from L and X?
F is just L sorted

	0	1	2	3	4	5	6	7	8	9	10
F =	a	a	a	a	a	b	b	c	d	r	r
L =	r	d	a	r	c	a	a	a	a	b	b

Note that L is the first column of A^s and A^s is in the same order as A.

If i is the k-th x in F then $T[i]$ is the k-th x in L.

Decoding Example

0 1 2 3 4 5 6 7 8 9 10

$$F = \begin{matrix} a & a & a & a & a & b & b & c & d & r & r \end{matrix}$$
$$L = \begin{matrix} r & d & a & r & c & a & a & a & a & b & b \end{matrix}$$
$$T = \begin{matrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ 2 & 5 & 6 & 7 & 8 & & & & & & \end{matrix}$$

Decoding Example

	0	1	2	3	4	5	6	7	8	9	10
$F =$	a	a	a	a	a	b	b	c	d	r	r
$L =$	r	d	a	r	c	a	a	a	b	b	

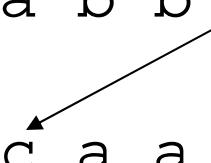
$T =$	0	1	2	3	4	5	6	7	8	9	10
	2	5	6	7	8	9	10				

Decoding Example

0 1 2 3 4 5 6 7 8 9 10

$F = \begin{matrix} a & a & a & a & a & b & b & c & d & r & r \end{matrix}$

$L = \begin{matrix} r & d & a & r & c & a & a & a & a & b & b \end{matrix}$



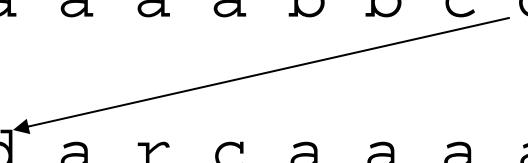
$T = \begin{matrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ 2 & 5 & 6 & 7 & 8 & 9 & 10 & 4 \end{matrix}$

Decoding Example

0 1 2 3 4 5 6 7 8 9 10

$F =$ a a a a a b b c d r r

$L =$ r d a r c a a a a b b



$T =$ 0 1 2 3 4 5 6 7 8 9 10

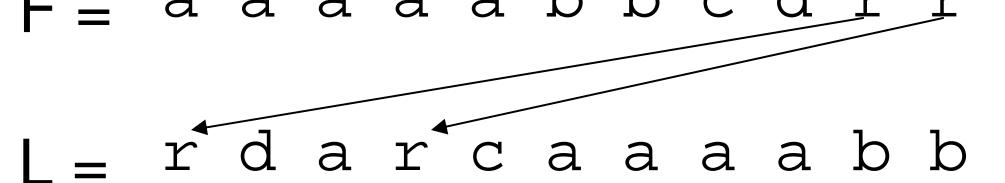
2 5 6 7 8 9 10 4 1

Decoding Example

0 1 2 3 4 5 6 7 8 9 10

$F = \begin{matrix} a & a & a & a & a & b & b & c & d & r & r \end{matrix}$

$L = \begin{matrix} r & d & a & r & c & a & a & a & a & b & b \end{matrix}$



$T = \begin{matrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 \\ 2 & 5 & 6 & 7 & 8 & 9 & 10 & 4 & 1 & 0 & 3 \end{matrix}$

Notes on BW

- Alphabetic sorting does not need the entire cyclic shifted inputs.
 - Sort the indices of the string
 - Most significant symbols first radix sort works
- There are high quality practical implementations
 - Bzip
 - Bzip2 (seems to be w/o patents)

Encoding Exercise

Encode the string abababababababab = $(ab)^8$

1. Find L and X
2. Do move-to-front coding of L.
3. Estimate the length of the code using first order entropy.

Decoding Exercise

Decode $L = \text{baaaaaba}$, $X = 6$

1. First Compute F and T
2. Use those to decode.