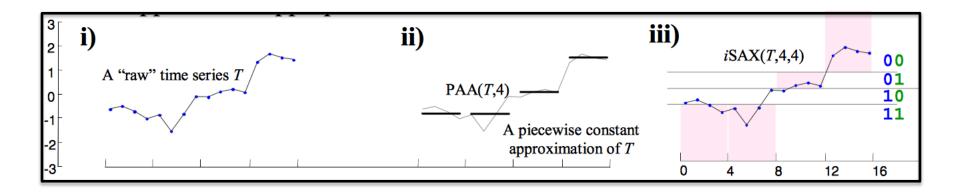
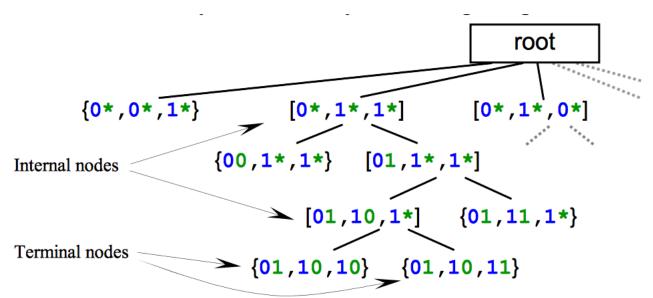
- A time series T of length n can be converted into a SAX representation
 - approximates T using a smaller number of segments (e.g. w=4)
 - each segment is represented by a discrete number
 - result is a SAX word (i.e. a vector) such as {11, 11, 01, 00}



- The SAX representation can be used as an index
 - Consider e.g. fixed cardinality of 8, word length of 4
 - An example T may map to $\{6^8, 6^8, 3^8, 0^8\}$
 - data for all Ts that can be represented by this SAX word can be stored in the same text file on disk (e.g. with name 6.8_6.8_3.8_0.8.txt)
- Problem: storage imbalance
- <u>Solution</u>: introduce a threshold for the number of time series that can be stored in a single file
 - If an insertion would cause threshold to be exceeded, split the file

- The diagram below illustrates an iSAX index as a tree
 - root node: represents complete SAX space
 - terminal node: leaf node containing pointer to file on disk
 - SAX word as index, contents are the actual time series data
 - internal node (new): designates split in SAX space
 - created when number of entries in a terminal node exceeds threshold

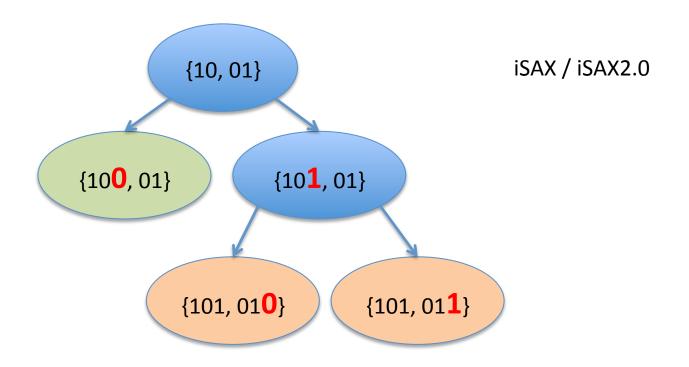


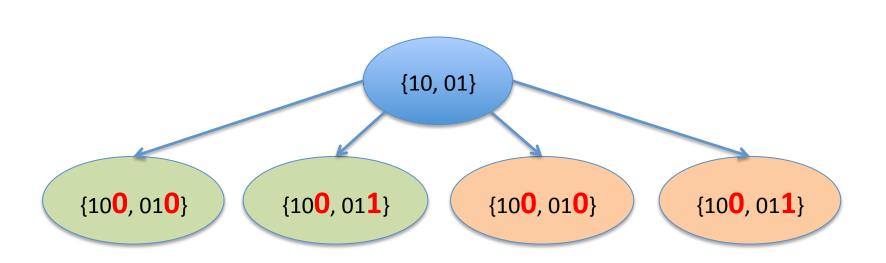
Observations:

- binary splits are along one dimension (sequentially)
 - creates two new words of increased cardinality
- "new" node splitting policy (iSAX 2.0) purports to provide better balance by determining optimal dimension
 - checks whether mean value is close to a breakpoint

However:

- balancing problems still exist
- no justification given for binary splits
 - note that root is connected to multiple nodes

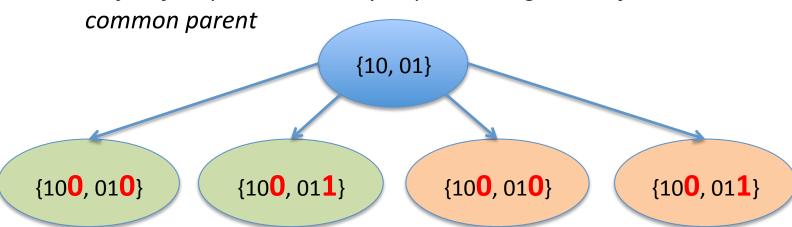




- We implemented iSAX tree as a "true" n-ary tree
 - splits series in a "full" terminal node into up to 'n' terminal nodes all located at the same depth in the tree
 - achieves better balance, faster traversals
- Class methods:
 - insert, delete, preorder (x2), similarity search (find_nbr)

```
randts.preorder() # get structure with counts
root
--->['10', '01', '10', '01']: 0
---->['100', '011', '100', '011']: 0
---->['1000', '0111', '1000', '0111']: 4
---->['1000', '0110', '1001', '0111']: 2
---->['1000', '0111', '1000', '0110']: 1
---->['1001', '0111', '1000', '0110']: 3
---->['100', '011', '100', '010']: 4
--->['01', '01', '01', '10']: 0
---->['011', '011', '011', '100']: 5
---->['011', '011', '011', '101']: 1
--->['10', '10', '01', '10']: 0
---->['100', '100', '011', '100']: 0
---->['1000', '1001', '0110', '1000']: 2
---->['1000', '1000', '0110', '1000']: 3
---->['1000', '1000', '0111', '1000']: 1
---->['1000', '1000', '0110', '1001']: 1
---->['100', '100', '010', '100']: 2
```

- Similarity search is "approximate"
 - intuition is that two similar time series are often represented by the same iSAX word
 - natural clustering
 - "ties" broken by computing Euclidean distance (but only for neighbors)
 - adjust for sparser nodes by implementing search for series with



- Also tested functionality on stock data
 - 1 year historical closing prices for S&P 500 stocks

```
stock.preorder ids() # see clusters of time series
---->['1101', '1011', '0010', '0101']: 2 ['LEG', 'PEP']
---->['1101', '1010', '0010', '0101']: 1 ['UPS']
---->['110', '100', '001', '011']: 4 ['FTI', 'PLL', 'TER', 'TMO']
---->['110', '101', '000', '011']: 0 []
---->['1101', '1010', '0001', '0110']: 2 ['GNW', 'HON']
---->['1101', '1010', '0001', '0111']: 1 ['HD']
---->['1101', '1011', '0001', '0110']: 2 ['PH', 'TGT']
---->['1100', '1011', '0001', '0111']: 1 ['VFC']
---->['1100', '1010', '0001', '0111']: 1 ['XRX']
---->['110', '101', '000', '010']: 3 ['KIM', 'RF', 'STI']
---->['110', '101', '001', '011']: 1 ['MDP']
--->['00', '11', '01', '01']: 1 ['AIG']
--->['11', '10', '00', '00']: 0 []
---->['111', '101', '001', '001']: 0 []
---->['1110', '1011', '0011', '0010']: 5 ['AIV', 'EQR', 'HAS',
---->['1110', '1010', '0011', '0010']: 4 ['CLX', 'SLE', 'SPG',
---->['1110', '1011', '0010', '0011']: 3 ['HOT', 'LTD', 'UNP']
---->['1110', '1010', '0010', '0011']: 3 ['LXK', 'VAR', 'WAT']
---->['110', '101', '001', '001']: 5 ['BXP', 'CBG', 'GPC', 'MAR', 'NKE']
```

