Sentiment Analysis for Dualcore Products

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Business Question

What changes should be made to Dualcore's product line after considering customer product reviews?

Solution: Sentiment Analysis

- Turn words into numbers
- Match with word sentiment dictionary
- Calculate opinion measures

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Sentiment = #positive - #negative range of -1 to 1 total count
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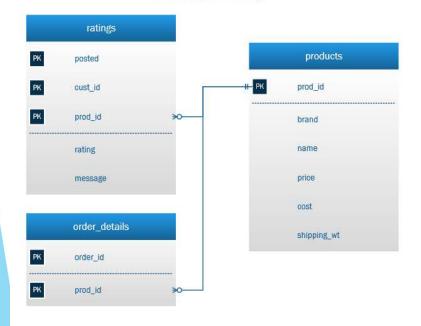
Overview: Tools

- ► Hadoop store and process in parallel of key-value data
 - Scoop ingest
 - HDFS store chunked data
 - MapReduce, Hive, Impala process
- Microsoft Access complex queries of dimensional data

Overview: Data Sources

Products, Order_Details and Ratings from Dualcore

DualCore (source files)



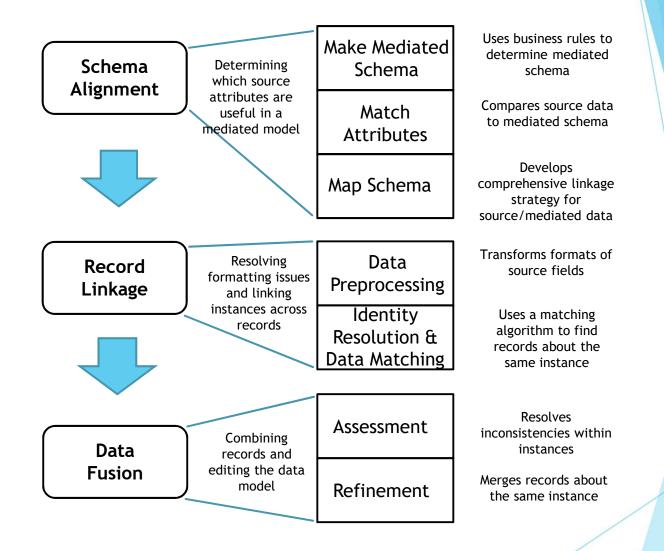
Word Dictionaries from Stanford and Cloudera

Positive Words:

good|happy|fantastic|great|perfect|enjoy|love|excellent|re
commend|high|well|best|excellent|better|highly|satisfied|b
argain|enjoy|awesome|pleased|enjoys|wellbuilt|decent|fin
e|great|bargain|easy|trust|perfectly|ideal|reliable|reasona
ble|worth|fine|flawless|sound|excelling|adept|supurb|impe
ccable|excelling|matchless|faultless|choicest|firstrate|incom
parable|terrific|outstanding|unparalleled|highest|unequaled
|ecstatic|excited

AND NOT not | don't | didn't

Data Integration Process



Data Integration Step 1: Schema Alignment

- Determine grain to answer business question (product)
- Determine attributes
- Locate data sources
- Map attributes from data sources
- Issues encountered:
 - Determine a grain which reflects magnitude of reviews
 - Dictionary (Stanford/movies, Cloudera/software version)

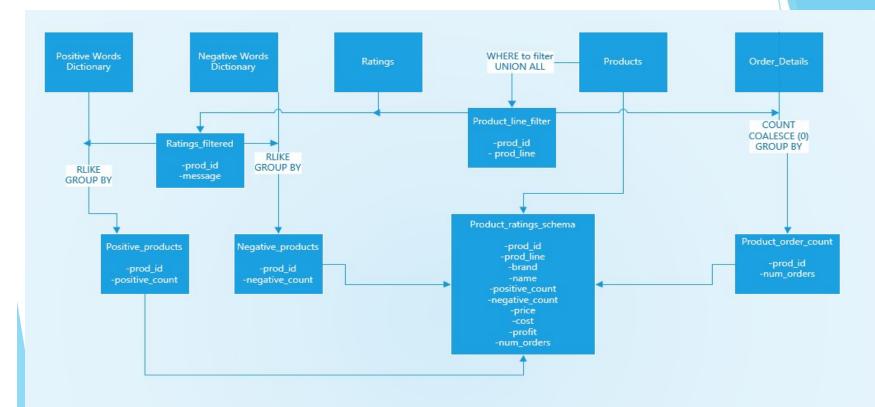
Mediated Schema Map

| | | | , | | | |
|----------------|--------------------------|------------------------------------|--|-------------------|---|---|
| 1 | products | order_details | ratings | dictionary | Example | Comments |
| prod_id | prod_id | prod_id | prod_id | Self. | 1273641 | Joins tables |
| prod_line | name CREATE FILTER | | | 5 | Tablet | Laptop, Server, Tablet |
| brand | brand | is . | | 30 | Byteweasel | |
| name | name | | | | Tablet PC (10 in. display, 16GB) | Filter to jus include Tablet PC |
| positive_count | | | messages AGGREGATED DEFAULT 0 | Positive words | 6 | Counted against dictionary. Aggregated by product |
| negative_count | | | Messages AGGREGATED DEFAULT 0 | Negative words | 2 | Counted against dictionary. Aggregated by product |
| sentiment | | | word count CALCULATED DEFAULT 0 | 81 | .7 | (Pos-neg) / (pos+neg) |
| positivity | | | word count CALCULATED DEFAULT 50 | | 80 | (Pos / (postneg) * 100 |
| price | price FORMATTED | | | | 495.79 | Convert to dollars |
| cost | cost FORMATTED | | | | 397.43 | Convert to dollars |
| profit | price-cost CALCULATED | | | | 98.36 | Calculate. Convert to dollars |
| num_orders | prod_id | prod_id CALCULATED DEFAULT 0 | | | 4 | Join tables on prod_id then COUN |

Data Integration Step 2: Record Linkage

- Formatting source attributes
 - prod_line, price, cost, profit, messages
- Transitioning rows to final grain
 - positive_count, negative_count, num_orders
- Issues encountered:
 - Scope/filter: tablets → tablet, server, laptop lines
 - Implement dictionary format
 - Attribute complexity with too much information
 - Aggregation, counts required intermediate tables

Intermediate Tables



- Further manipulation within Access:
 - Calculate sentiment and positivity measurements
 - Group by prod_line, brand, counts

Data Integration Step 3: Data Fusion

- Consider incongruencies
- Identifying true values
- Issues encountered:
 - ▶ Handing Nulls: counts to 0, opinions to neutral (0 or 50).
 - Implementing Nulls: COALESCE and IF
 - Limitations of complex joins and aggregations (dimensional)

- Analyzed by Product Line
 - Laptop only Acme cooling pads, highest laptop only neutral
 - Server -Megachamp Lemmon have high sales but low ratings. Olde-Grey highly rated. All motherboards poorly rated
 - Tablet drop Orion and United Digital. Byteweasel sells the most and has highest positivity, but we are loosing \$1 per sale.

| Product Line | sentiment | positivity | #ratings | brand | name | # orders | profit |
|--------------|-----------|------------|----------|-----------|-------------------------|----------|----------|
| Laptop | -1 | 0 | 1 | XYZ | Laptop Cooling Pad | 33 | \$26.16 |
| | -0.1 | 44 | 9 | ElCheapo | Laptop Cooling Pad | 3653 | \$3.08 |
| | 0 | 50 | 0 | Duff | Laptop Cooling Pad | 0 | \$11.73 |
| | 0 | 50 | 0 | TPS | 1GB DDR2 667 Laptop R | 226 | \$12.09 |
| | 0.1 | 56 | 9 | Sparky | 2GB DDR2 800MHz Lapt | 3484 | \$15.27 |
| | 0.2 | 58 | 12 | SuperGami | 1GB DDR2 667 Laptop R | 5148 | \$10.76 |
| | 0.5 | 75 | 57 | ACME | Laptop Cooling Pad | 4789 | \$0.45 |
| | 1 | 100 | 5 | Overtop | 1GB DDR2 667 Laptop R | 3639 | \$8.18 |
| Server | -1 | 0 | 6 | Megachan | 4 TB NAS Server | 2391 | \$108.96 |
| | -1 | 0 | 2 | Orion | 1 TB NAS Server | 0 | \$46.52 |
| | -0.3 | 33 | 6 | Lemmon | 8 TB NAS Server | 1556 | \$51.82 |
| | -0.3 | 36 | 11 | Orion | Server Motherboard | 4932 | \$13.95 |
| | -0.1 | 46 | 24 | TPS | Server Motherboard | 6635 | \$20.82 |
| | 0 | 50 | 0 | BuckLogix | Server (1U rackmount, I | 0 | \$465.36 |
| | 0 | 50 | 0 | BuckLogix | Server (1U rackmount, I | 0 | \$885.60 |
| | 0 | 50 | 0 | Dorx | 1 TB NAS Server | 0 | \$20.80 |
| | 0 | 50 | 0 | Dualcore | Home Media Server | 0 | \$8.84 |
| | 0 | 50 | 0 | Dualcore | Server (1U rackmount, I | 0 | \$589.32 |
| | 0 | 50 | 0 | Gigabux | Server (2U rackmount, | 0 | \$583.76 |

- Analyzed by Brand
 - 5 brands elicit strong negative opinions
 - ▶ Bigdeal 2 products total, brand sentiment of -0.6, high sales
 - United Digistuff 3 products, brand sentiment of -0.37, high sales

| brand - | Avg Of sentiment - | Avg Of positivity - | Avg Of total_pos_neg - | Sum Of num - | CountOfproc - |
|------------------|--------------------|---------------------|------------------------|--------------|---------------|
| Bigdeal | -0.60 | 20 | 6 | 2520 | |
| United Digistuff | -0.37 | 32 | 23 | 6162 | 3 |
| XYZ | -0.33 | 33 | 1 | 69 | |
| Lemmon | -0.15 | 42 | 8 | 2156 | 2 |
| Orion | -0.15 | 43 | 61 | 21015 | 10 |
| TPS | -0.05 | 48 | 24 | 6861 | 2 |
| ElCheapo | -0.05 | 47 | 15 | 4416 | 2 |
| Megachango | -0.02 | 49 | 69 | 11615 | 6 |
| Dorx | 0.00 | 50 | 0 | 0 | 1 |
| Dualcore | 0.00 | 50 | 0 | 0 | 2 |
| Duff | 0.00 | 50 | 0 | 0 | 1 |
| Gigabux | 0.00 | 50 | 0 | 0 | 1 |
| Krustybitz | 0.00 | 50 | 0 | 0 | 3 |
| Yoyodyne | 0.00 | 50 | 0 | 0 | |
| Homertech | 0.15 | 57 | 13 | 4584 | 2 |
| ACME | 0.17 | 58 | 57 | 4827 | 3 |
| SuperGamer | 0.20 | 58 | 12 | 5148 | 1 |
| Byteweasel | 0.20 | 60 | 215 | 124245 | 2 |
| BuckLogix | 0.27 | 63 | 8 | 1939 | 3 |
| Sparky | 0.28 | 64 | 17 | 7115 | 4 |
| Tyrell | 0.28 | 64 | 12 | 4252 | 5 |
| Olde-Gray | 0.50 | 75 | 18 | 7743 | |
| Electrosaurus | 1.00 | 100 | 4 | 954 | 2 |

- Analyzed by Counts
 - ▶ Top 7 counts, all but one had better than average reviews
 - Motherboards have high number of reviews, and they are negative

| total_pr - | positivity - | sentime + | positi\ - | negativ - | brand + | name + | prod | num_orders - |
|------------|--------------|-----------|-----------|-----------|-------------------------|---------------------------------|--------|--------------|
| 203 | 62 | 0.2 | 126 | 77 | Byteweasel | Tablet PC (10 in. display, 64 G | Tablet | 119801 |
| 57 | 75 | 0.5 | 43 | 14 | ACME | Laptop Cooling Pad | Laptop | 4789 |
| 44 | 75 | 0.5 | 33 | 11 | Megachango | Sleeve for Tablet - Blue | Tablet | 6714 |
| 24 | 46 | -0.1 | . 11 | 13 | TPS | Server Motherboard | Server | 6635 |
| 19 | 68 | 0.4 | 13 | 6 | Megachango | Sleeve for Mini Tablet - Black | Tablet | 2454 |
| 17 | 71 | 0.4 | 12 | 5 | Orion | Tablet PC (10 in. display, 32 G | Tablet | 5642 |
| 17 | 76 | 0.5 | 13 | 4 | Olde-Gray | Home Media Server | Server | 5814 |
| 15 | 33 | -0.3 | 5 | 10 | United Digistuff | Tablet PC (10 in. display, 64 G | Tablet | 3246 |
| 12 | 58 | 0.2 | 7 | -5 | SuperGamer | 1GB DDR2 667 Laptop RAM | Laptop | 5148 |
| 12 | 58 | 0.2 | 7 | 5 | Byteweasel | Tablet PC (10 in. display, 16 G | Tablet | 4444 |
| 11 | 36 | -0.3 | 4 | 7 | Orion | Server Motherboard | Server | 4932 |
| 11 | 64 | 0.3 | 7 | 4 | Homertech | Tablet PC (7 in. display, 16 GE | Tablet | 3156 |
| 9 | 22 | -0.6 | 2 | 7 | Orion | Tablet PC (10 in. display, 64 G | Tablet | 3848 |
| 9 | 44 | -0.1 | . 4 | 5 | ElCheapo | Laptop Cooling Pad | Laptop | 3653 |
| 9 | 56 | 0.1 | . 5 | 4 | Sparky | 2GB DDR2 800MHz Laptop RA | Laptop | 3484 |
| 8 | 12 | -0.8 | 1 | 7 | United Digistuff | Tablet PC (7 in. display, 16 GE | Tablet | 2916 |
| 8 | 88 | 0.8 | 7 | 1 | BuckLogix | Tablet PC (7 in. display, 8 GB) | Tablet | 1939 |
| 7 | 57 | 0.1 | . 4 | 3 | Orion | Tablet PC (7 in. display, 32 GE | Tablet | 1024 |
| 7 | 86 | 0.7 | 6 | 1 | Orion | Tablet PC (10 in. display, 32 G | Tablet | 3141 |
| 6 | 0 | -1 | . 0 | 6 | Megachango | 4 TB NAS Server | Server | 2391 |
| 6 | 33 | -0.3 | 2 | 4 | Lemmon | 8 TB NAS Server | Server | 1556 |
| 6 | 50 | 0 | 3 | 3 | ElCheapo | Mobile Bluetooth Keyboard a | Tablet | 763 |
| | | | | | | | | |

- Negative NGRAMS
 - mediocre, not great, 30+
 - bad, horrible, poor, 2x

```
{"ngram":["not","great","but","not"],"estfrequency":35.0}
```

{"ngram":["great","but","not","bad"],"estfrequency":35.0}

{"ngram":["i","think","it","is"],"estfrequency":32.0}

{"ngram":["think","it","is","mediocre"],"estfrequency":32.0}

{"ngram":["we","hate","this","item"],"estfrequency":2.0}

{"ngram":["item","was","a","bad"],"estfrequency":2.0}

{"ngram":["was","a","bad","value"],"estfrequency":2.0}

{"ngram":["this","item","was","the"],"estfrequency":2.0}

{"ngram":["this","product","was","horrible"],"estfrequency":2.0}

{"ngram":["item","was","a","poor"],"estfrequency":2.0}

Positive NGRAMS

- decent, 25x
- happy, better than

```
{"ngram":["i","feel","it","is","decent"],"estfrequency":25.0}
```

{"ngram":["this", "is", "a", "decent", "product"], "estfrequency":19.0}

{"ngram":["i","am","very","happy","with"],"estfrequency":9.0}

{"ngram":["this","is","better","than","the"],"estfrequency":8.0}

{"ngram":["better","than","the","previous","model"],"estfrequency":6.0}

{"ngram":["is","better","than","the","previous"],"estfrequency":6.0}

{"ngram":["very","happy","with","the","product"],"estfrequency":5.0}

{"ngram":["am","very","happy","with","the"],"estfrequency":5.0}

{"ngram":["i", "absolutely", "recommend", "this", "product"], "estfrequency":4.0}

{"ngram":["very","happy","with","this","item"],"estfrequency":4.0}

Future Steps ...

- Sentiment of phrases
- Scale of positive/negative
- Further train dictionary

Questions?