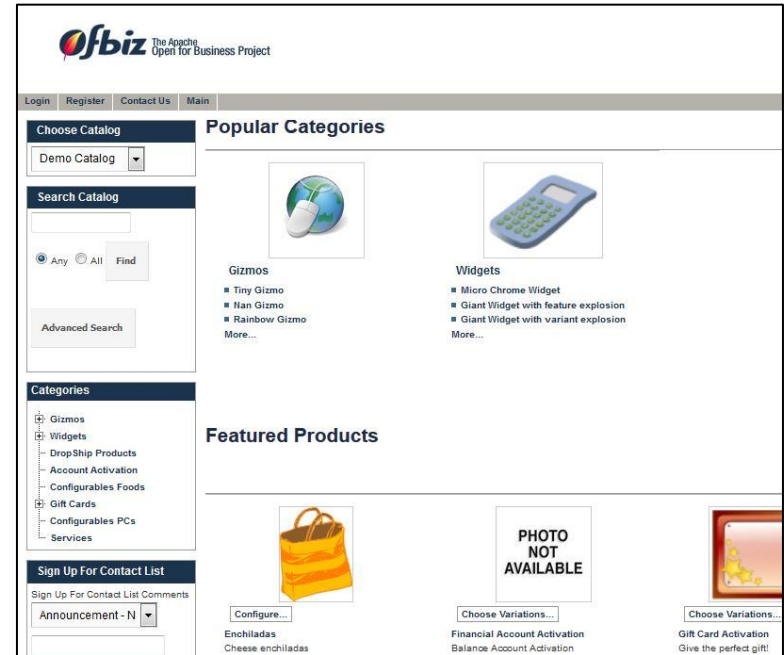


Introduction to a Recommender System for Apache OFBiz™

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– “Apache OFBiz™ is an open source product for the automation of enterprise processes that includes framework components and business applications for ERP (Enterprise Resource Planning), CRM (Customer Relationship Management), **E-Business / E-Commerce**, SCM (Supply Chain Management), MRP (Manufacturing Resource Planning), MMS/EAM (Maintenance Management System/Enterprise Asset Management), POS (Point Of Sale).”



OFBiz demo e-commerce site

– Currently OFBiz doesn't include any service that generates Recommendations for online shoppers

Outline

- **Introduction**
 - Recommendations
 - Product Association Recommendations
- **A Product Association Recommender for OFBiz**
 - Out of the box support for Product Association
 - Recommender based approach to Product Association
 - Prototype of a Product Association Recommender
 - Demo
- **From the prototype to a production-ready component**
 - Summary of the main prototype features
 - Improvements and automation strategies

Introduction

Introduction

Recommendations (1/2)

- **Context**

- **System** (e-commerce site, web-radio, etc.) **that provides a service** (selling, streaming, etc.) **on some products** (books, music, films, etc.)
- **Set of users' that interact with the system** (browse, rate, click, etc.) **and utilize the service** (buy a book, listen to a song, etc.)

- **Recommendation**

- Information given to the users to support their choices (suggestions about products to buy, etc.)
- Based on
 - Users' preferences (preferred literary genre, preferred music genre, etc.)
 - Historical data (about past purchases, products' ratings, etc.)

Introduction

Recommendations (2/2)

• Examples of Recommendations

– **Personalized:** takes into account historical data or specific preferences of the user who is receiving the recommendation

- Collaborative filtering (User-user, Item-item, Dimensionality reduction)
- Content-based filtering


– **Non personalized:** does not take into account historical data or preferences of the user who is receiving the recommendation

- Summary statistics (aggregated opinion recommenders)
- **Non personalized Product Association Recommendation** (*Ephemerally personalized:* takes into account the current activity of the user, e.g., the product that is being checked out or is being viewed, etc.)

Recommended for You
(to, click here.)

These recommendations are based on items you own and more.


view: **All** | [New Releases](#) | [Coming Soon](#)

1.  **Daredevil: Born Again**
by Frank Miller (January 20, 2010)
Average Customer Review: ★★★★★ (42)
In Stock

List Price: \$19.99
Price: \$12.64
62 used & new from \$8.64

I own it Not interested ★★★★★ Rate this item

Recommended because you purchased **Batman: Year One** and more (Fix this)




★★★★★ 701

FREE Shipping on orders of \$35 or more

Product Features
... Simple and versatile, this watch is perfect for everyday wear.

[Clothing, Shoes & Jewelry](#)



★★★★★ 173

FREE Shipping

Product Features
Package includes: 1 x Blue Smartwatch, 1 x Black Rubber Band, 1 x Black Casual Watch

[Clothing, Shoes & Jewelry](#)

Customers Who Bought This Item Also Bought



Black Casual Watch
★★★★★ (308)
\$8.30



Watch with Black Rubber Band
★★★★★ (154)
\$8.30

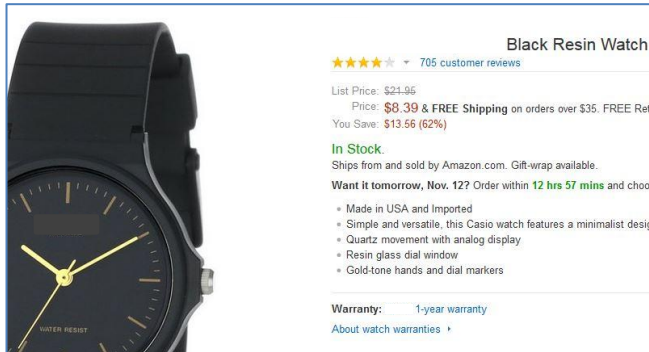


Watch with Black Band
★★★★★ (117)
\$8.99

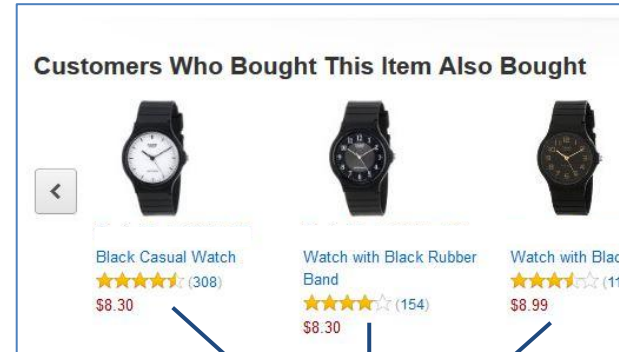
Non personalized Product Association Recommender (1/4)

- **Key concept**

- People who did X (bought some product, listened to a song, etc.) also did Y



↓
People who bought X...



... also bought Y

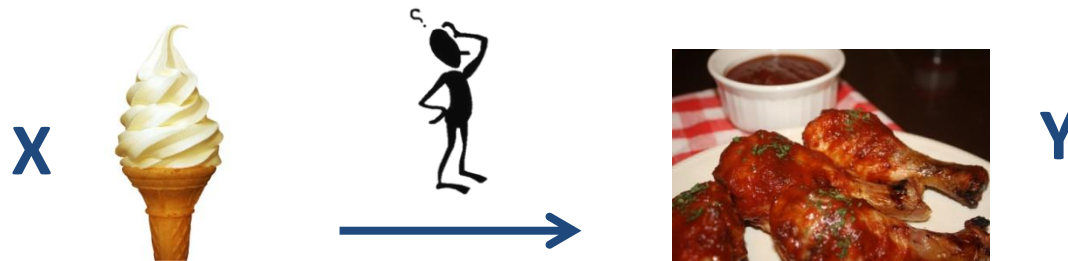
- **Key question**

- Given X, what are the products Y whose purchase is related ('associated') to the purchase of X?

Non personalized Product Association Recommender (2/4)

- **Naïve algorithm**

- **X**: product for which we want Product Association Recommendations
- Recommend Y if most of the people who bought X also bought Y
 - **Maximization of $P[Y|X]$** (read: probability of Y given X)
- **Drawback** : recommend popular products that are not related to X
- **Example**
 - **X**: ice-cream
 - **Y**: popular barbecue sauce
 - Most of the people who bought the ice-cream also bought the barbecue sauce → **high $P[Y | X]$** → **recommend the barbecue sauce**



Non personalized Product Association Recommender (3/4)

- **Proper algorithm**

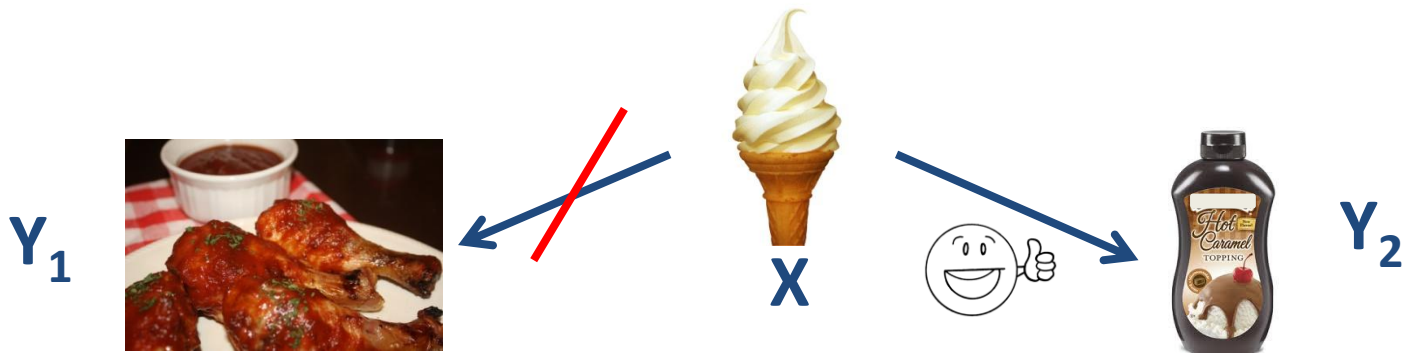
- **X**: product for which we want Product Association Recommendations
- **Recommend Y if the purchase of X increases the probability of buying Y**
 - $P[Y|X] > P[Y|\sim X]$ (read: probability of Y given not X)

Non personalized Product Association Recommender (4/4)

- **Proper approach**

- **Example**

- **X**: ice-cream
- **Y₁**: popular barbecue sauce; **Y₂**: caramel topping for ice-creams
- **Y₁**: most of the people who bought the ice-cream also bought the barbecue sauce → **high P[Y₁ | X]**, but also most of the people who have never bought an ice-cream bought barbecue sauce → **same high P[Y₁ | ~X]**
 - **P[Y₁ | X]=P[Y₁ | ~X] → Y₁ NOT recommended**
- **Y₂**: most of the people who bought the ice-cream also bought the caramel topping → **high P[Y₂ | X]**, but very few people among those who have never bought an ice-cream bought a caramel topping → **low P[Y₂ | ~X]**
 - **P[Y₂ | X] > P[Y₂ | ~X] → Y₂ recommended!**



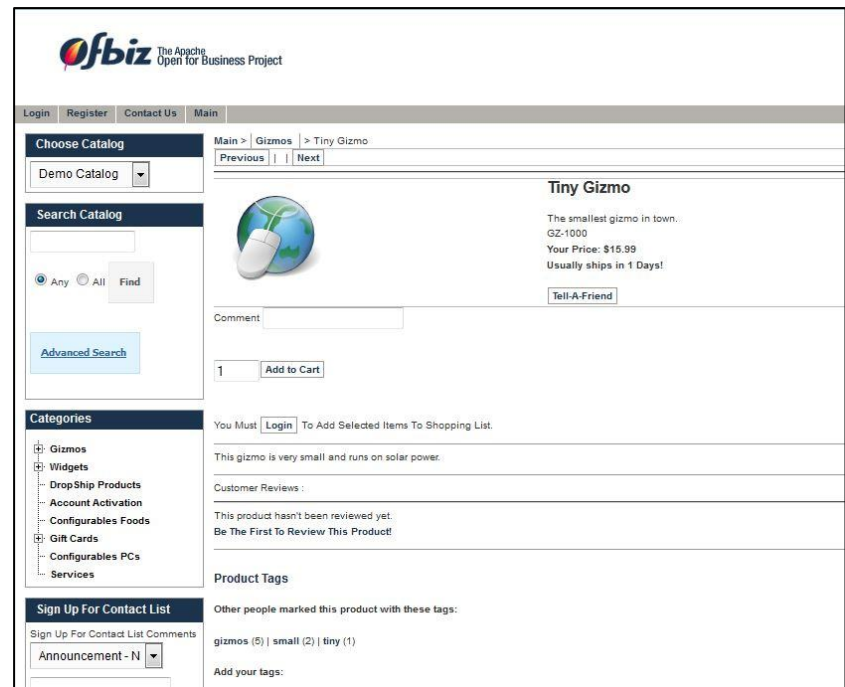
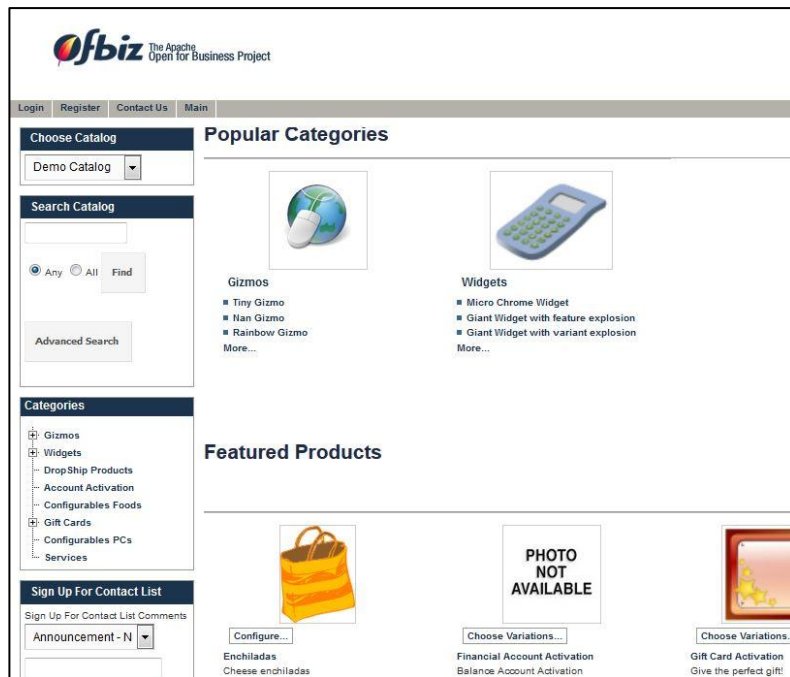
A Product Association Recommender for Apache OFBiz

A Product Association Recommender for OFBiz

Out of the box support for Product Association (1/2)

- Demo ecommerce site
(<http://localhost:8080/ecommerce/>)

- Product detail page



A Product Association Recommender for OFBiz

Out of the box support for Product Association (2/2)

– Manual insertion of a record in the ProductAssoc entity with productAssocTypeId set to “ALSO_BOUGHT”

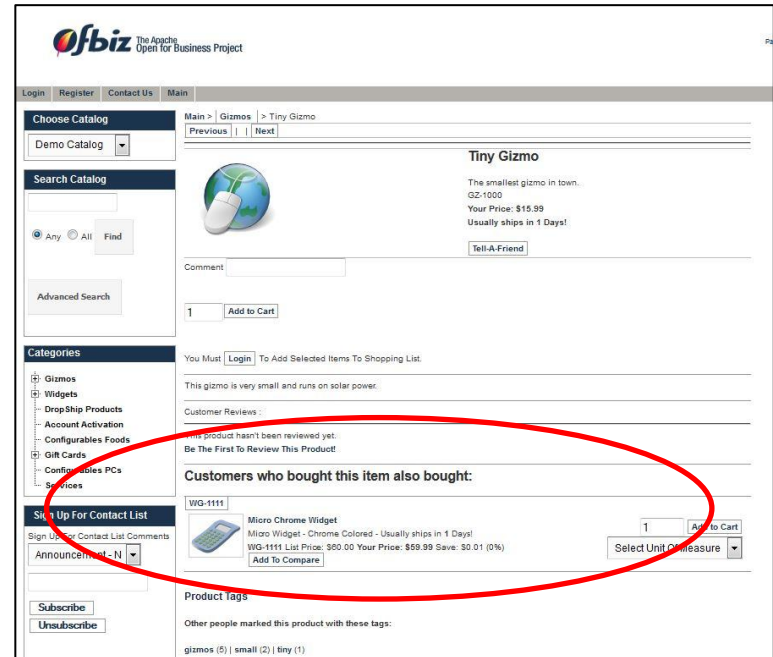
<entity-engine-xml>

```
<ProductAssoc  
  productId="GZ-1000"  
  productIdTo="WG-1111"  
  productAssocTypeId="ALSO_BOUGHT"  
  fromDate="2014-09-02 12:00:00.000"/>
```

</entity-engine-xml>

• Drawbacks

- Requires manual maintenance
- Not based on any Product Association recommender algorithm



A Product Association Recommender for OFBiz

Recommender based approach

• Procedure

1) Generation of associations based on Product Association algorithm

- **X** : product for which we want Recommendations
- **recommend Y if the purchase of X increases the probability of buying Y**

$$P[Y|X] > P[Y|\sim X]$$

2) Storage of associations in ProductAssoc entity

• Implementation of Product Association algorithm

– **Y**: items bought together with X in the same orders

– **Computation of $P[Y|X]$ and $P[Y|\sim X]$**

– $\hat{P}[Y|X] = (\# \text{ of orders containing } Y \text{ and } X) / (\# \text{ of orders containing } X)$

– $\hat{P}[Y|\sim X] = (\# \text{ of orders containing } Y \text{ but not } X) / (\# \text{ of orders not containing } X)$

– **Assessment of $P[Y|X] > P[Y|\sim X]$**

– $\hat{P}[Y|X] > \hat{P}[Y|\sim X]$ and

– Log-likelihood ratio hypothesis test

– Null hypothesis $P[Y|X]$ equal to $P[Y|\sim X]$

A Product Association Recommender for OFBiz

Prototype of a Product Association Recommender (1/2)

- **“Recommender” component** (<https://cwiki.apache.org/confluence/x/OYPfAg>)
 - **getRecommendedAssociatedProducts** Java service
 - **Input**
 - product ID (**X**) for which we want to provide Recommendations
 - N max number of recommended products to return
 - **Output**
 - list of recommended products for X
 - **Main steps:**
 - 1) Finds the products ordered together (items in the same sales orders) with X
 - 2) For each ‘co-ordered’ product Y, computes the conditional probabilities and collect the products that pass the test
 - 3) Returns the first N (highest scores) recommended products and stores them in ProductAssoc entity (productAssocTypeId = “ALSO_BOUGHT”)



A Product Association Recommender for OFBiz

Prototype of a Product Association Recommender (2/2)

– **createTestDataForProductAssociationRecommender** Java service: creates sales orders that simulate purchases with known conditional probabilities

- Input

- N number of orders

- Products

- LED TV (X)

- USB PEN DRIVE (Y_1) ($P[Y_1 | X] = P[Y_1 | \sim X]$)

- TV AUDIO SYSTEM (Y_2) ($P[Y_2 | X] > P[Y_2 | \sim X]$)

- Conditional probabilities

- $P[\text{LED TV}] = 0.5$

- $P[\text{TV AUDIO SYSTEM} | \text{LED TV}] = 0.7$; $P[\text{TV AUDIO SYSTEM} | \sim \text{LED TV}] = 0.1$

- $P[\text{USB PEN DRIVE} | \text{LED TV}] = 0.9$; $P[\text{USB PEN DRIVE} | \sim \text{LED TV}] = 0.9$

A Product Association Recommender for OFBiz

Demo (1/2)

- Install the "Recommender" component under the hot-deploy folder
- Load demo products (ant load-demo)
- Run OFBiz (%OFBIZ_HOME\tools\startofbiz.bat or %OFBIZ_HOME\tools\startofbiz.sh)
- Web Tools (<http://localhost:8080/webtools/control/main>) -> Run Service

1)

The screenshot shows the OFBiz Service Engine interface. The breadcrumb navigation is Applications > Framework Web Tools > Service Engine. The tabs are SERVICE REFERENCE, JOB LIST, THREAD LIST, SCHEDULE JOB, and RUN SERVICE. The 'Service' field contains 'createTestDataForProductAssoc' and the 'Pool' field contains 'pool'. A 'Submit' button is visible at the bottom.

The screenshot shows the 'Schedule Job' configuration for the 'createTestDataForProductAssoc' service. The 'numberOfOrders (Integer)' field is set to '1000' and is marked as 'Required'.

2)

The screenshot shows the OFBiz Service Engine interface. The breadcrumb navigation is Applications > Framework Web Tools > Service Engine. The tabs are SERVICE REFERENCE, JOB LIST, THREAD LIST, SCHEDULE JOB, and RUN SERVICE. The 'Service' field contains 'getRecommendedAssociatedProd' and the 'Pool' field contains 'pool'. A 'Submit' button is visible at the bottom.

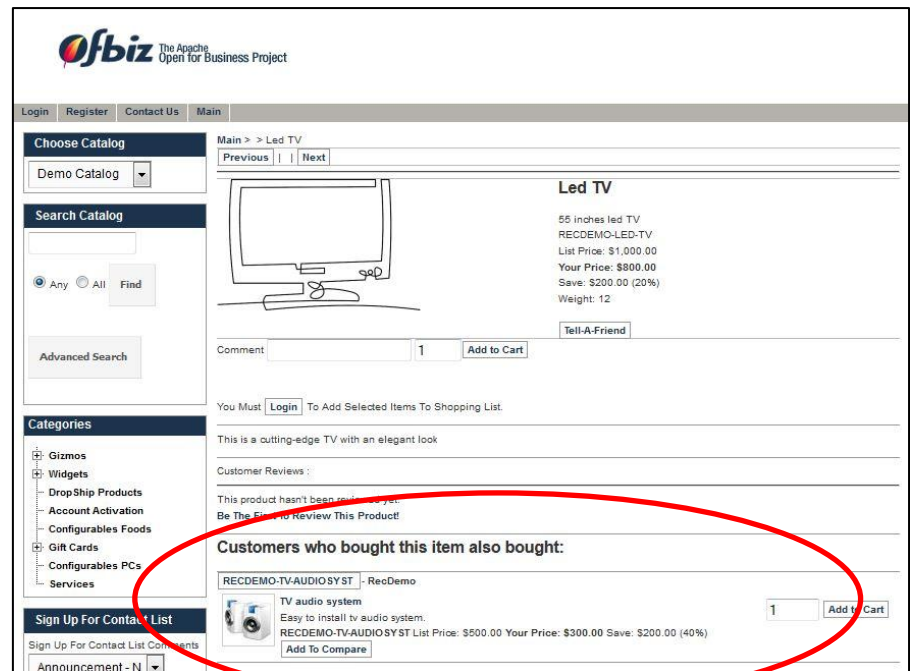
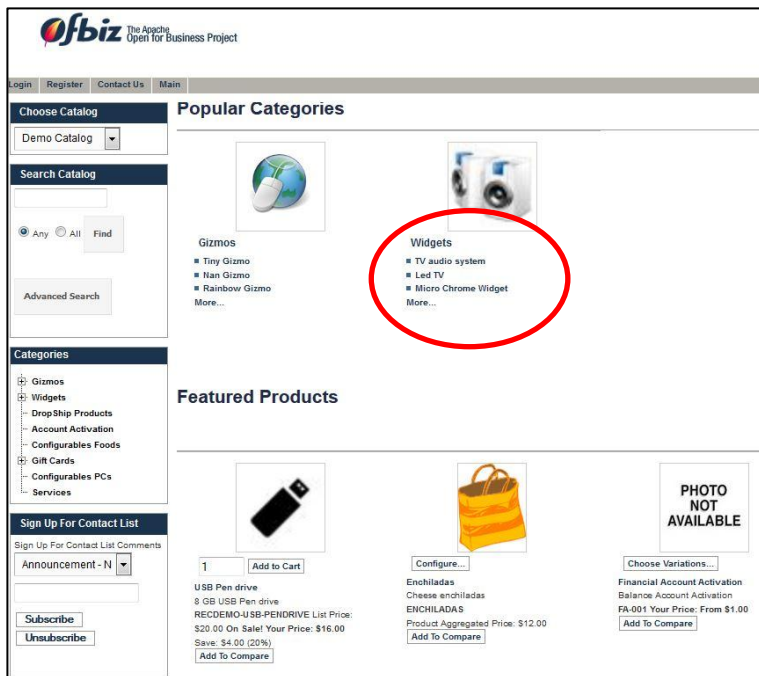
The screenshot shows the 'Schedule Job' configuration for the 'getRecommendedAssociatedProd' service. The 'numbOfRecommendedProducts (Integer)' field is set to '1' and the 'productId (String)' field is set to 'RECDEMO-LED-TV', both marked as 'Required'.

A Product Association Recommender for OFBiz

Demo (2/2)

- Demo ecommerce site
(<http://localhost:8080/ecommerce/>)

- Led TV detail page



From the prototype to a production-ready component

From the prototype to a production-ready component

Summary of the prototype main features

- ‘Recommender’ hot-deploy component for the generation of Product Association Recommendations (downloadable from <https://cwiki.apache.org/confluence/x/OYPfAg>)
- Services
 - getRecommendedAssociatedProducts
 - createTestDataForProductAssociationRecommender
 - removeRecommenderTestData
- Products ‘bought together’ \leftrightarrow in the same order
- Product Association algorithm based on computation of conditional probabilities and log-likelihood statistical hypothesis testing

From the prototype to a production-ready component Improvements and automation strategies

- Introduction of a configurable time-window for considering products as “bought together”
- Bulk processing (all products, categories, etc.)
- Load-testing (on hundreds of products and thousands of orders) and performance improvements
- Automation strategies
 - service triggered, e.g., by the creation of an order (e.g., ECA)
 - scheduled job (every night, week, etc.)
- Integration of business rules
- **Community feedback and contributions are very welcome!**

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About me

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