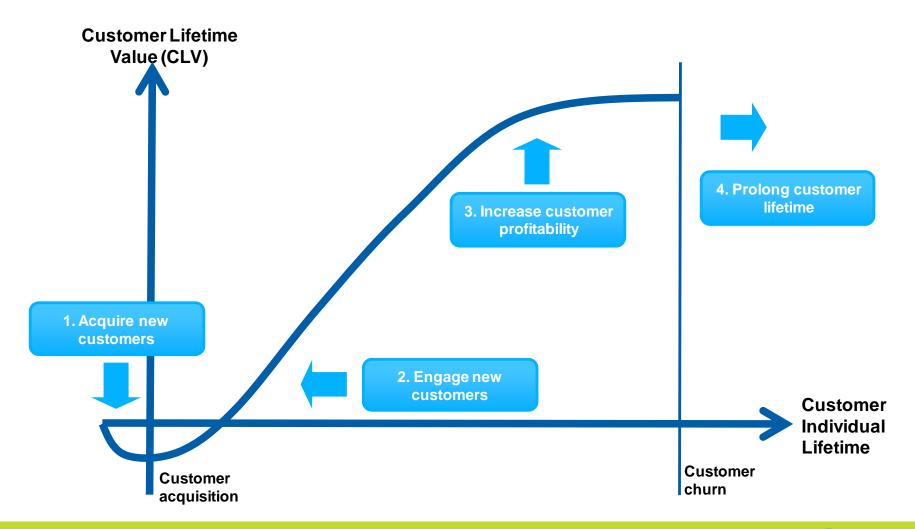


Customer Retention and CHURN Prevention Comptel Social Links

Diego Becker - SVP CALA Matti Aksela - VP Analytics May 8th, 2012

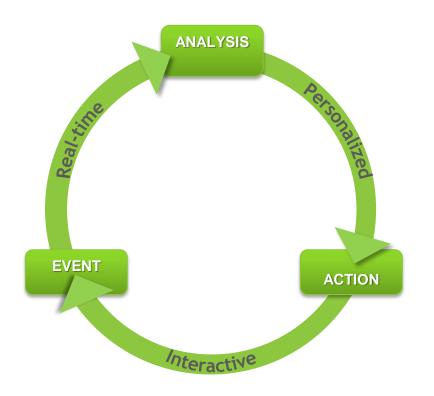
Maximising customer value





What is Intelligent Customer Interaction?

...as an input to realtime analytics engine..



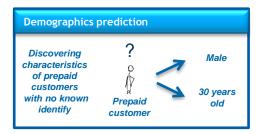
..to drive realtime actions

Real-time network events..

Analytics Securing Revenue

Example use cases

Marketing insight



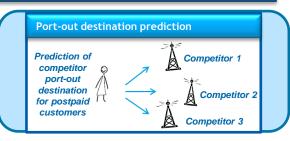




Revenue stimulation and churn management

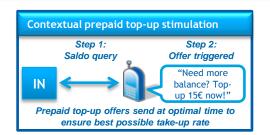






Contextual customer engagement









Value Propositions

Delivering unmatched marketing results

"Advanced analytics drives new revenue growth and reduced churn"

Best-in-benchmark analytics accuracy

"Predictive modelling, social network analysis and machine learning are key features of our analytical models"

Profit from Big Data

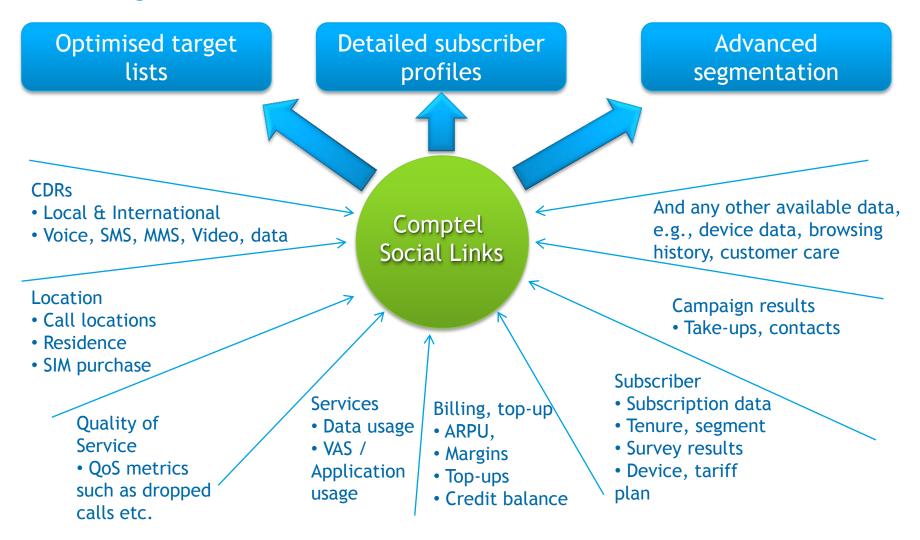
"No need to restrict the amount of Big Data, the more you have, the better we are"

Fast time to revenue

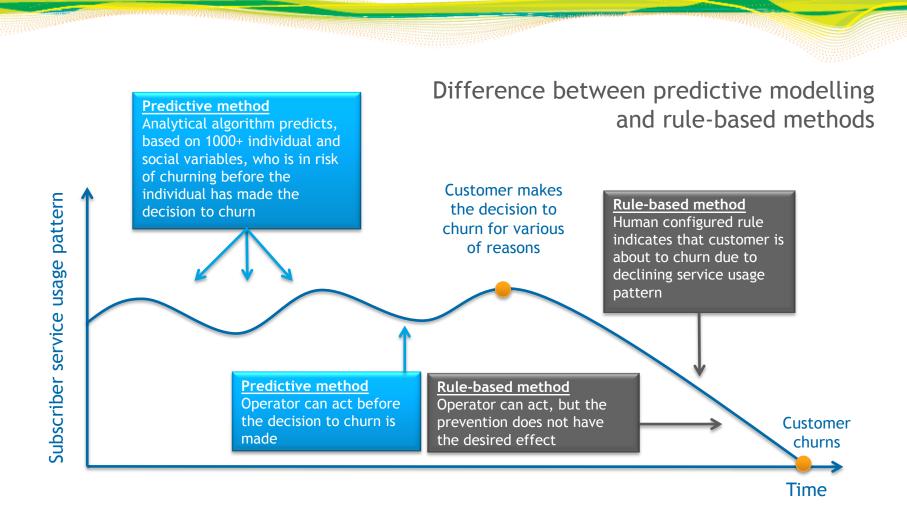
"When data is available, our operational algorithms will turn it into new revenue in eight weeks"



Solution Can Use Any Data Source Depending on the Use Case Targets



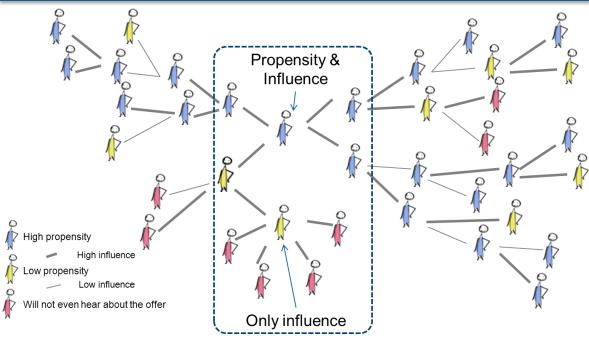
Predictive Modelling Provides Unique Insights into Future Customer Behavior





Powered by Social Network Analysis

Social network and alpha influencers



Algorithm creates social network models and indicates who are the influential alpha users in these networks



Using changes in social networks as an input, algorithms make accurate predictions about individual's behavior



Case study: Proactive Pre-paid Churn Prevention

Background

Location: Eastern Europe

ARPU: 12€
Pre-paid: 85%
Mobile subs: 5 million
Churn rate: 7% monthly
Campaigns: Monthly

Target group: Top 10% monthly

Before Social Links

High monthly churn rate: 7%

Campaigns did not have retention impact

Monthly campaigns were unprofitable

Solution

Step 3. Optimisation of personalised retention offer for each high churn risk customer and contacting customer with SMS offer

Step 2. Uplift modelling to optimise revenue from Churn L. Step 4. Prodictions

Step 1. Prediction of likely churn point

Active stage of customer life-cycle

Results

More retained customers and secured revenue

Enabled by granular campaign optimisation and predictive targeting



21% reduction in churn rate 25% increase in revenue



Case study: Prediction of Post-paid Churners' Port-out Destination

Background

Location: **APAC ARPU:** 61€ Pre-paid: 53%

Mobile subs: 4.8 million Churn rate: 1-2% monthly

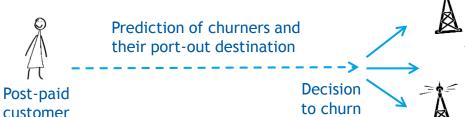
Before Social Links

Lack of competitive insight and port-out destinations

No intelligence or no tools to prevent churn to specific competitor

Inefficient retention campaigns

Solution



Port-out to competitor x

Port-out to competitor y

Port-out to competitor z

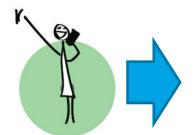
Results

12

High quality input data

High quality modelling

Excellent prediction



87% correctly predicted port-out destination for the target group!



Case study: Pre-paid Top-up Optimisation

Background

Location: MEA ARPU: 13€ Pre-paid: 80%

Mobile subs: 15 million
Churn rate: 1.5% monthly

Campaigns: Monthly Target group: 1 million

Before Social Links

Sub-optimal revenue from prepaid top-up campaigns

Top-up recharge rewards always fixed, e.g., 5% of the required action

No analytics used for optimising top-up offers

Solution

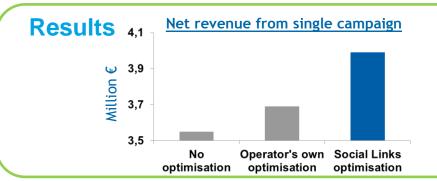
Identify those likely to respond positively

Tailor personalised required actions

Tailor personalised top-up rewards

"Top-up 15€ now, get 2€ extra!"

Individual offer to all subscribers





400 000€ more revenue from a campaign vs. status quo

threefold improvement on operator's own method



13

