

Managing uncertainty in animal planning

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CONFIDENTIAL AND PROPRIETARY Any use of this material without specific permission of McKinsey & Company is strictly prohibited Over the years, Supply Chain management has evolved in depth and granularity to meet increased market and production complexity



Operational logistics to ensure supply of production lines and delivery to customers **SC as function nonexistent**, logistics reporting to sales and manufacturing Adoption of S&OP - Integrated planning with M&S and supply Limited planning interactions (monthly) Single direction flow – ensure alignment Low visibility and granularity Creation of **Control Towers** – intra-month operational adjustments forums Increased frequency of interactions ~weekly Selected feedback loops Increased visibility

But this journey is far from over, and three global megatrends are pushing that Supply Chains evolve to yet a new standard



Rapidly changing of consumer demand and behaviors are <u>forcing deep revision of operations strategy and setup</u>





Rural Growth

Wealth and growth moving to regions that have not been served in the past



Demanding customer expectations

Growing service expectations combined with shorter lead time to fulfill customer orders





Individualization

Large increase in SKU range due to ever growing customization driving complexity in supply chain



Multitude of options

Consumers have options to buy from **multiple channels** creating need for more granular demand and supply synchronism

In an environment of increasingly complexity, companies need to setup their supply chain to deliver More granularity Higher precision

Better integration

Massive advancements in technology and innovation are prime to enable the digitization of Operations





Data, computational power, and connectivity

Significantly reduced costs of computation, storage, and sensors Reduced cost of small-scale hardware and connectivity (IoT)



Analytics and intelligence

Breakthrough advances in artificial intelligence and machine learning Improved algorithms and improved data availability



Human machine interaction

Quick proliferation via portable interfaces Breakthrough of optical head-mounted displays



Digital-to-physical conversion

Advances in artificial intelligence, machine vision, M2M communication, and cheaper actuators

Expanding range of materials, rapidly declining prices for printers, increased precision/quality

Sensors in everything, networks everywhere automate anything, and analyze everything to significantly improve performance and customer satisfaction

The size of the prize – applying digital supply chain levers, huge potential can be unlocked in all supply chain categories



In addition to that, agricultural animal protein supply chain is even more complex due to three main challenges they have to deal with



Disassembly Trade-offs

Once volume is locked, complexity is still high since all parts must be sold, and decisions of specific cuts will affect the availability of others



3



Demand fluctuates in terms of product mix due to consumer preferences, and total volume due to events at the global economy level, such as trade wars and market openings and closures



Long Supply Chain

Having a long supply chain implies less flexibility and higher costs to adjust production throughout the process

1 There are key disassembly decisions to be made that makes it <u>complex and requires analytics to maximize profits</u>



Key decisions involved

How to define the best volumemarket mix

How to decide which plant will serve which market (some plants can only supply to specific markets due to sanitary restrictions)

How to cope with the cuts that are not so demanded and are perishable

How to maximize revenues and margins with all these constraints



How to optimize?

Use analytics to maximize expected margin

- Historical data (internal)
- Internal plans and targets
- Production constraints (capacity, licenses, etc.)
- External data (competitors and market)
- Promotion/clearance effect (price elasticity)
- Cannibalization effect (internal)
- Etc.

B Map alternative solutions for output overflow

- Create new brands
- Expand to new markets (local and/or foreign)
- Create new products (i.e. fresh product line)

Consider the cost and value added by each raw material to create each SKU

American cuts



British Cuts



2 Changes in demand occur often not only in the product mix due to <u>consumer preferences, but also due to events at the global economy</u>

U.S.-China Trade War



2018: China imposed tariffs (25%) on pork imported from the U.S.

Sep/2019: China rolling back tariffs on pork imports from the U.S. (limited to a certain amount)





3 In long supply chains, continuous investment progressively increase the <u>cost of changing volume, eventually locking-in total production levels</u>





The complete cycle from beginning until first processed animal ~ 940 days

There is a large spectrum of digital drivers and levers that help <u>companies deal with these challenges and evolve to a new standard</u>





One approach is to adopt an ongoing demand refinement process, with the support of a control tower for short-term adjustments

PO **Build demand** forecast **Refine demand** ("unrestricted") (semi-restricted) **Detail demand** There is still a small scope to (restricted) adjust total volume, but **Control Tower** Overall volume is settle with no overall capacity is settled (minor adj) scope for adjustment Destination markets. Volume is given and there is no Last opportunities to adjust weight branding, packaging, etc. are scope to adjustment group distribution still adjustable Short-term Branding, packaging and final management of Remaining levers can still be Cuts and weight group destination markets should be demand/ adjusted distribution is still adjustable defined in this steps production shortage to a certain extent or excess

Flexibility to adjust demand

For demand forecasting, machine learning led to 40% decrease in stockout, 35% less inventory and 13pp better forecast accuracy

Case 1

Context

Personal care direct selling company in LatAm facing pressure to increase EBITDA

Inaccuracy on forecast as the root-cause for more than two-thirds of shortages

High complexity driven by:

Frequent promotions and high depth of discounts

Large and discontinuous portfolio (~40% of portfolio removed and re-introduced every month)

Constant innovation

Many **qualitative factors driving demand not codified in databases**; planners manually identify and evaluate those factors to estimate demand



Approach

End-to-end planning program including:

- Product segmentation and strategy
- Inventory optimization
- S&OP review and Control tower implementation
- Demand forecasting solution

In Demand forecasting, we have created a **hybrid** solution combining machine learning and human judgement to achieve the best estimate, using...

- External data (macroeconomic indicators) and internal data (historical demand, regular and discounted prices, promotion types, product exposure in ads, among others)
- Demand forecasting unit (DFU):
 - SKU group (one level above SKU)
 - Promotion type (e.g., straight discount, bundle)
 - Sales period
- Features to deal with demand behaviors:
 - Client's purchase behavior
 - Cannibalization
 - Pantry loading



Shortage savings:

Full potential estimated to decrease stockout in 40% (30% already captured in 6 months)

Accuracy improvements:

Monthly forecast accuracy in DFU: +13pp better than baseline (forecast error -32%)

Inventory reduction of 35%

A hybrid solution combined the best of human emotional judgement with advanced analytics' empirical approach...

How the hybrid solution is structured

Company: human emotional solution produced by planners

PROS

- Uses different sources of qualitative information
- Able to adjust forecast based on qualitative non-codified information
- More updated on recent trends
- Sees intangible features on sales vehicles

CONS

- Depends on different experiences of multiple people
- Uses non-standard criteria to perform same task



MACHINE LEARNING: structured, mathematical solution PROS

- Able to "see" nuances of dozens of variables based on thousands of past experiences
- Follows a clear logic
- Able to adjust based on every new datapoint included every campaign
- Able to learn "demand behavior" from other products

CONS

Based uniquely on what is available on databases

PROS

More robust to inaccuracies on either side (lower variance in error) Higher accuracy

CONS

Depends on the maintenance of current process with additional steps

...and uses 5 models to: predict demand, decide how to blend it w/ human's manual prediction, and correct outliers and negative bias

How the multiple models in the hybrid solution are structured



Improvements were created by the understanding of outliers and creation of multiple models and multiple features

History of the development



Importance of features in the model

Regression for demand-price Average demand Normalized discounted price Regression for demand-ad weight Dif % of price to re-seller Ad page weight Promo "Volume discount" type? Promo "Buy quantity N..."? Max incentive offered to re-Seller No. of months since Launch Number of concepts on Ad Seasonality factor Max % incentive offered to re-Seller Other 170 variables

Most features contain a link with price/promo elasticity, which is not seen in a typical demand forecasting problem

An optimization of the planning process using Llamasoft Supply Chain <u>Guru for Logistics Footprint can reduce shipping cost by more than 5%</u>

Case 2

Context



Lack of a structured planning process that allows adjustments to occur frequently in a centralized manner

Inability to see all bottlenecks at local levels of the system due to lack of granularity and integration of inputs

Decision to allocate logistics flows not optimally due to absence of analytical support



Approach

Designed and implement end-to-end planning process with multiple horizons

Developed analytical tool using the Llamasoft Supply Chain Guru software to optimize volume allocation in the footprint physical and temporal flow of products

Optimization has the objective function to maximize product prices and minimize costs incurred in the chain (origination, processing, storage, transportation)

The tool also includes constraints such as asset capacity (processing, receiving and dispatch), origination volumes, market demand, modal transport contracts more than 750 restrictions

A customized set of dashboards was developed to evaluate results and drive decisions

Inputs Supply Chain By Determ Supply Chain By Determ Unit of the second second

Front End (preview)



Optimization model implementation that translates the complexity of the footprint and granularity required for decision making

Optimization of 500,000 decision points through 1.2 million of variables, for example:

- Unbalance of bottlenecks in origination
- Need for new investments in assets
- Market to be served
- Transport contracts and ports

Ongoing to deliver more than 5% reduction in shipping costs

Many organizations are shifting to a digital management of their Supply Chains through intelligent dashboards and predictive analytics to accelerate decision making



Performance outlook based on current performance, automated root cause and predictive analyses allows anticipation of decisions, minimizing costs and disruptions



To quickly and easily analyze all results generated by the Solution, McKinsey created its Always On Control Tower solution



Using Always On can increase significantly the quality analysis on forecast

Summary of how to transform into a digital supply chain

1 Define your digital supply chain journey

Define the baseline of your starting point Understand digital waste, e.g. by conducting a digital walkthrough Layout future state, but be flexible in execution

2 Get the right digital talent, integrate and nurture

Develop talent internally or hire externally Install creativity processes – e.g. hack week, hackathon Understand and communicate changes in the organizational setup

3 Create the right environment for your digital transformation journey Allow to fail and ensure the right to stop and reprioritize Provide the right flexibility – e.g., via Amazon Web Services 2-speed architecture to conserve legacy and push agility Foster start-up to push innovative topics and re-integrated Start small, test fast and scale up



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Thank you