

Atelier 3: Computational and Mobil based solutions in personalized nutrition – effectiveness and limitations

Dr. David De Lorenzo, Universitat Pompeu Fabra

David de Lorenzo is a B.Sc. (Honours) graduated from the University of Navarre, and PhD (cum laude, Molecular Population Genetics) from the University of Barcelona. He has focused his scientific career on the understanding of the genetic basis of complex diseases, and the study of the interactions between genetic and nutritional factors in relationship to human health. His past professional experience consists of different positions at the University of Texas Health Science Center (USA), the Ludwig-Maximilians University of Munich (Germany), and the University of Lleida (Spain). Currently David de Lorenzo is associated professor at the university Pompeu Fabra in Barcelona (Spain), and director of the area of Personal Genomics at NIMGenetics SL (Scientific Park of Madrid, Spain). He is member of the Spanish Society of Genetic Counselling (SEAGEN), the Spanish Association of Human Genetics (AEGH), and the Spanish Society of Genetics (SEG).

Modération: Dr. Robert Sempach, Migros Kulturprozent

Computational and mobile based solutions in personalised nutrition

Effectiveness and limitations

Dr. David de Lorenzo
Director - Personal Genomics Area

Associate Professor



IT solutions in personalised nutrition

1. Current state of the art
2. Why IT solutions for PN?
3. Limitations and Current Challenges
4. Ethics and Perspectives



State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Dietary Recalls
- Food Frequency Questionnaires (FFQ)
- Multiple-day Food Records
- Specific Food Questionnaires



State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Dietary Recalls

In-depth "unannounced" interview that collects detailed information on all foods and beverages consumed by a participant during the previous 24 hours.



State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Dietary Recalls
- Food Frequency Questionnaires (FFQ)

Report of the frequency of consumption and portion size of X items over a defined period of time.



State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Food Frequency Questionnaires (FFQ)



State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Food Frequency Questionnaires (eFFQ)

VioScreen/VioFFQ User's Guide

State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Dietary Recalls
- Food Frequency Questionnaires
- Multiple-day Food Records

A food record or food diary is a detailed description of all foods and beverages consumed over a period of three to seven days.



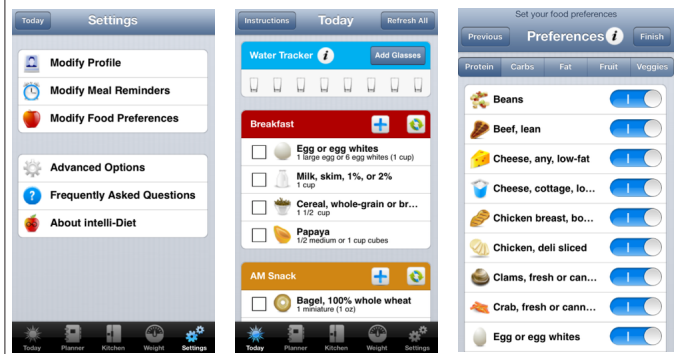
State of the art - IT solutions for Personalised Nutrition

1. Nutrition assessment

- Dietary Recalls
- Food Frequency Questionnaires
- Multiple-day Food Records: IT Tools
 - Web Food Diaries



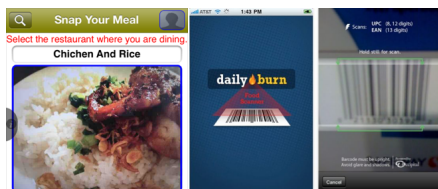
State of the art - IT solutions for Personalised Nutrition



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 - Food Scan Apps - Input: Photos, BarCodes, etc.



Why IT solutions for Personalised Nutrition?

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Why IT solutions for Personalised Nutrition?

Yatekomo Pollo
Marca: Gallina Blanca

Energía	Grasas	Grasas saturadas	Azúcares	Sal
282.00 kcal/1179.00 kJ	13.00 g	7.10 g	1.90 g	2.30 g
1414	19 %*	36 %*	2 %*	39 %*

190g	310.00g	% IRV/RN *
88.00 kcal/368.00 kJ	282.00 kcal/1179.00 kJ	1414 %
Grasas 4.00g	13.00g	19 %
de las cuales saturadas 2.30g	7.10g	36 %
Hidratos de carbono 11.00g	35.00g	
de los cuales azúcares 0.60g	1.90g	2 %
Fibra alimentaria 0.70g	2.30g	
Proteínas 1.60g	5.00g	
Sal 0.76g	2.30g	39 %

*Ingesta de Referencia/Valor de Referencia de Nutrientes de un adulto medio (8400 kJ/2000 kcal)

Número de raciones por envase: 1

Why IT solutions for Personalised Nutrition?

Food Scanner Horizon Prize

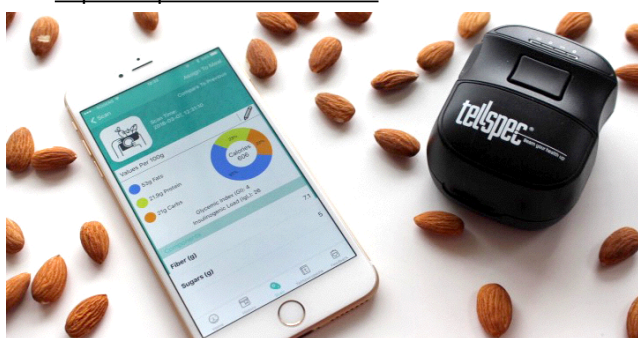
<https://www.youtube.com/watch?v=v0uggsj4Ars>

If labels are not enough...

Why IT solutions for Personalised Nutrition?

Tellspec

<http://tellspec.com/howitworks/>



Why IT solutions for Personalised Nutrition?

Scio

<https://www.consumerphysics.com/myscio/order/>



Why IT solutions for Personalised Nutrition?

Foodscan + Metabolic biomarkers:

Discover yourself

Cue is a deep health tracker that lets you go beyond tracking steps and your weight.

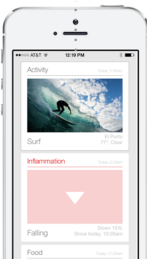
Easily access deeper information from your body.

Load a cartridge,
add a sample,
and get results on your phone.

Information is Power

Track your body at the molecular level

Proactively tune your diet and exercise with the help of Cue's Smart Recommendation so that you create your best story



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New Integrated Medical Genetics

Why IT solutions for Personalised Nutrition?

+ Metabolic biomarkers:

Normal	Borderline	High Need	Supplementation for High Need
Antioxidants			
Vitamin A / Carotenoids	Vitamin C		
Vitamin E / Tocopherols	α-Lipoic Acid		
CoQ10			
B-Vitamins			
Thiamin - B1			
Riboflavin - B2	Niacin - B3		
Pyridoxine - B6			
Biotin - B7			
Folic Acid - B9			
Cobalamin - B12			
Minerals			
Magnesium	Manganese		
Molybdenum		Zinc	Zinc - Dose = 30 mg
Vitamin D			
		Vitamin D	Vitamin D - Dose = 4,000 IU

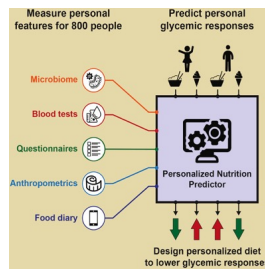
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NIMGenetics
New Integrated Medical Genetics

Why IT solutions for Personalised Nutrition?

1. Nutrition assessment
2. Personalised recommendations

Based on Nutrition assessment + metabolic & metagenomic markers



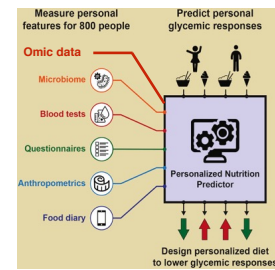
Source: Personalized Nutrition by Prediction of Glycemic Responses. Zeevi et al. Cell 163, 1079–1094, November 19, 2015

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Why IT solutions for Personalised Nutrition?

1. Nutrition assessment
2. Personalised recommendations

Based on Nutrition assessment + genetic (GxE), metabolic and metagenomic markers



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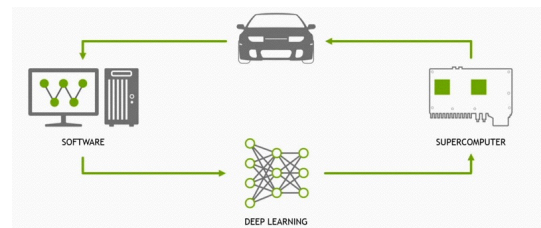


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Why IT solutions for Personalised Nutrition?

1. Nutrition assessment
2. Personalised recommendations
3. (Machine) Learning from Genetic background + Environment-Nutrition and Health Outputs



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IT for Personalised Nutrition is not one-directional knowledge, but bi-directional

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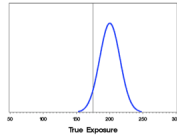
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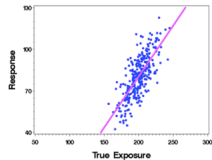
Limitations and Current Challenges

Two main current applications:

1. Describe intake distributions (define needs, both at population and individual level)



2. Estimate Diet-Health relationships



Limitations and Current Challenges

Limitations of FFQ

1. Cognitively challenging
2. Affected by recent diet
3. Finite food list
4. Lack of detail: assumptions required in converting to nutrient and food group intake

Limitations and Current Challenges

Limitations of FFQ

1. Cognitively challenging
2. Affected by recent diet
3. Finite food list
4. Lack of detail: assumptions required in converting to nutrient and food group intake
5. BUT Inexpensive, and very easy to implement through web applications.

Limitations and Current Challenges

Limitations of Dietary Recalls (24 h recalls)

1. Less cognitively challenging (relies on short- term recall)
2. Rich detail & fewer assumptions required in converting to nutrient and food group intake
3. Aims to capture recent diet -> Need more than one to assess usual intake
4. Expensive to collect and code

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4. Expensive to collect and code
5. BUT Computational approaches (web-based recalls) makes possible this approach at much lower cost.

Limitations and Current Challenges

Limitations of Food Records

1. Less cognitively challenging (does not rely on memory)
2. Aims to capture current diet (often over several consecutive days)
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4. Recording may affect intake (reactivity)
5. Expensive to code

Limitations and Current Challenges

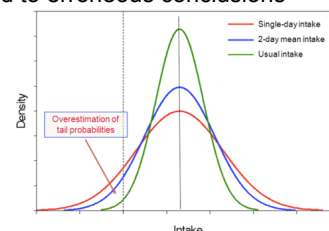
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4. Recording may affect intake (reactivity)
5. Expensive to code
6. BUT Computational approaches, such as food record apps on mobile phones that may reduce much of the manual coding required.

Limitations and Current Challenges

Challenges to estimate usual intake

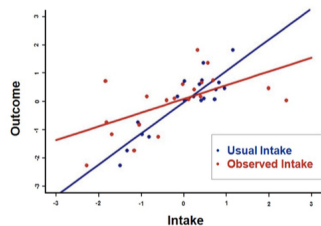
1. **Self-report** instruments used to assess usual dietary intake are affected by several types of measurement error – If we ignore this error, our results may be biased
2. Using **observed rather than true intake** can lead to erroneous conclusions



Limitations and Current Challenges

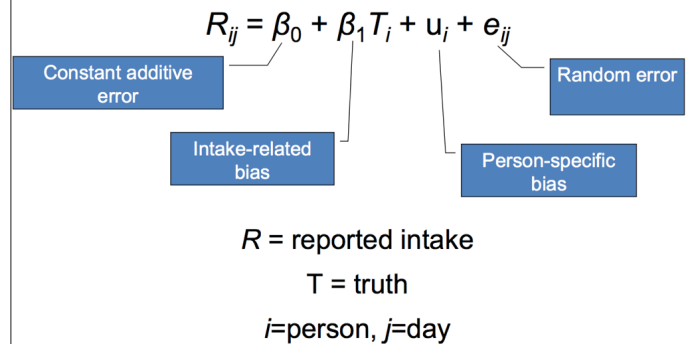
Challenges to estimate usual intake

- Regression using **observed rather than true intake** produces attenuated slope estimate and loss of power to detect relationship between a dietary exposure and an outcome



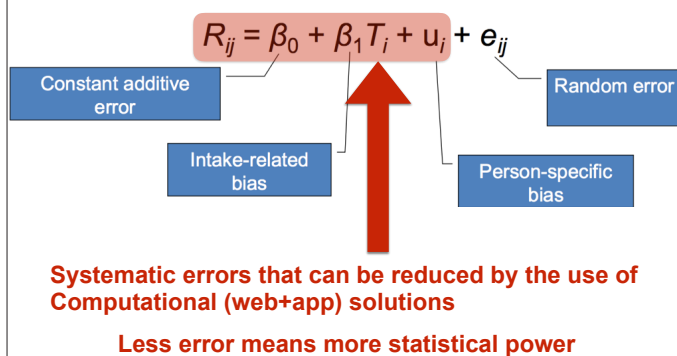
Limitations and Current Challenges

The structure of measurement error



Limitations and Current Challenges

The structure of measurement error



Perspectives & Ethics

Without -omic markers, use of IT-solutions provide:





- Better estimates of population intake requirements
- Better regression of nutritional factors and health effects at population level

With -omic markers, use of IT-solutions provide:

- Better estimates of individual intake requirements
- Better regression of nutritional factors vs health effects

Analysis of the interaction Nutrient intake, omic markers & other environmental values (by machine learning) allows the elaboration of precise personalised nutritional recommendations

Perspectives & Ethics

Huel® Nährwertinformationen Huel kaufen Alle Produkte Forum Mehr    

Alles, was dein Körper braucht

Das Nahrungsmittel der Zukunft

Von Experten entwickelt. Gratis Versand.

HUEL BESTELLEN

Perspectives & Ethics

Ethical considerations

1. Complex systems biology: Genes implicated in different diseases with unwanted information - ApoE.
2. The results may have implications not only for the person who is being analysed, but also for genetically related family, raising questions about the sharing of the information with the patient's families.
3. In addition, a genetic test is always a potential paternity test.
4. There is growing concern about the best way to ensure that the results of a genetic test be kept confidential, and how to avoid discrimination based on genetic information.

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