

Computational Gastronomy



Infosys Centre for
Artificial Intelligence

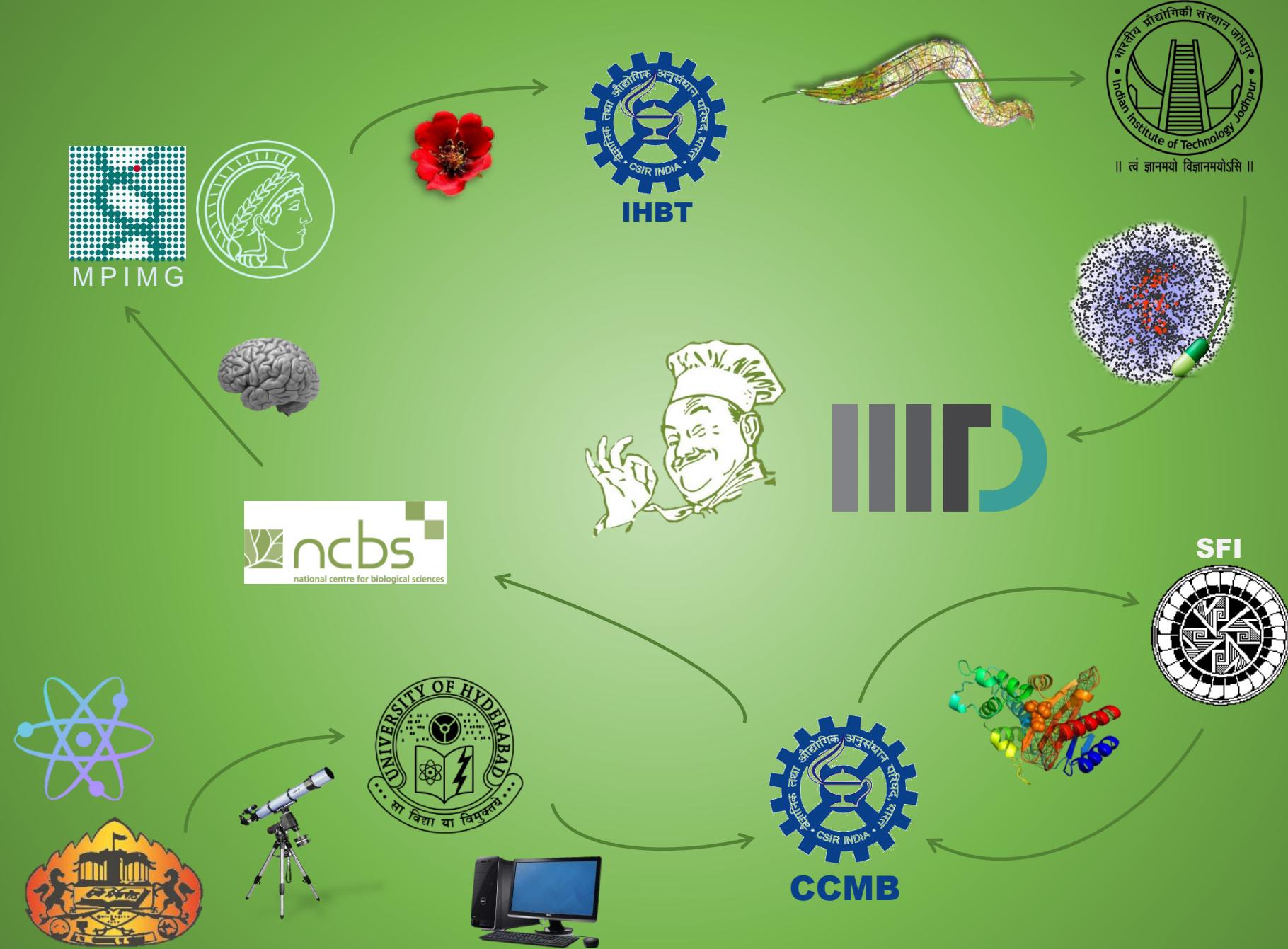
GANESH BAGLER



A Data-driven Science of Food, Flavors, Nutrition, Health & Sustainability



Computational Gastronomy = Food + Data Science + Artificial Intelligence



From astronomy to gastronomy



Why food?



Cooking is alchemy

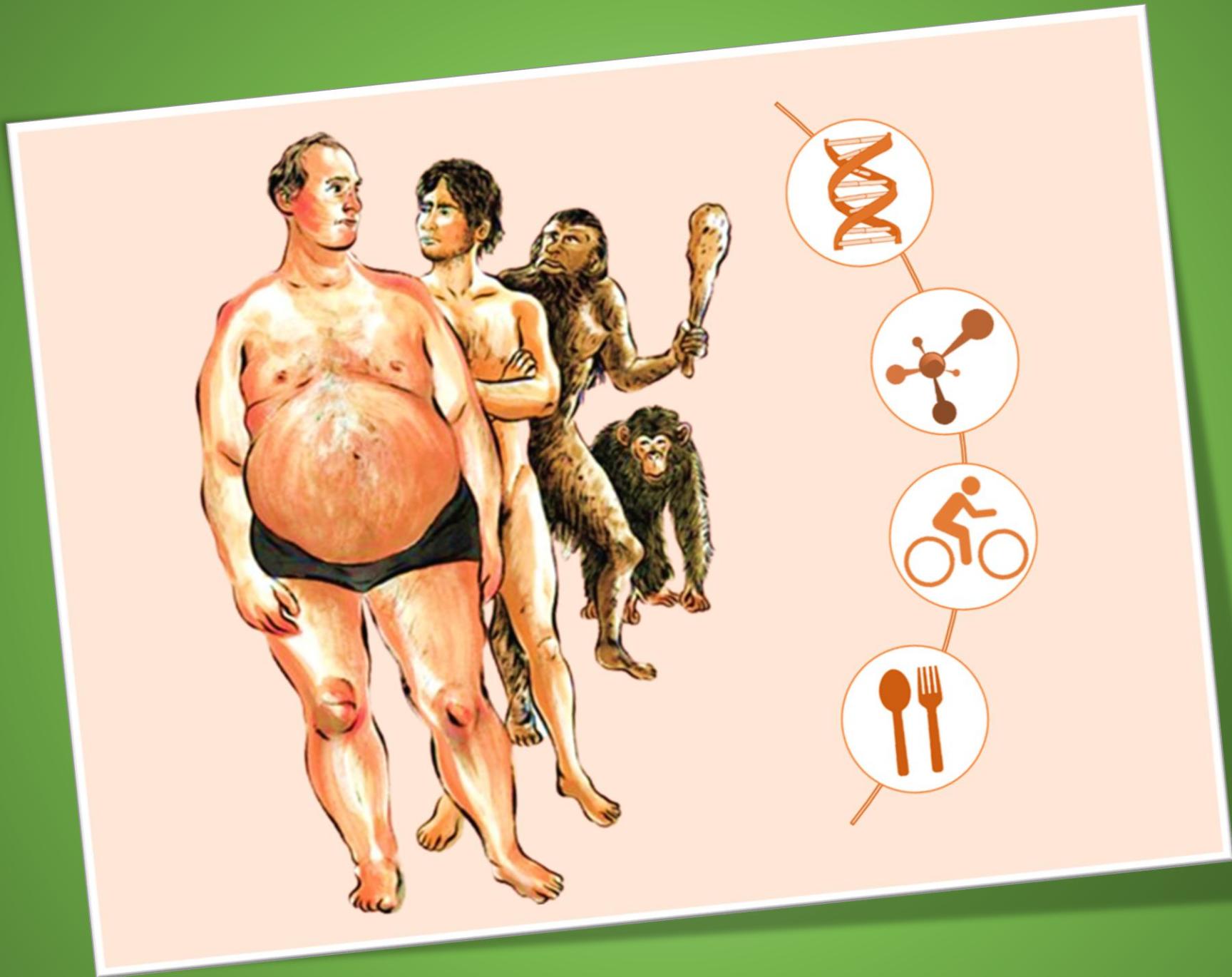
Being Human



Being Human



'Catching Fire—How cooking made us human' by Richard Wrangham





Innovation [noun]
make changes in something established,
especially by introducing new methods, ideas, or products.

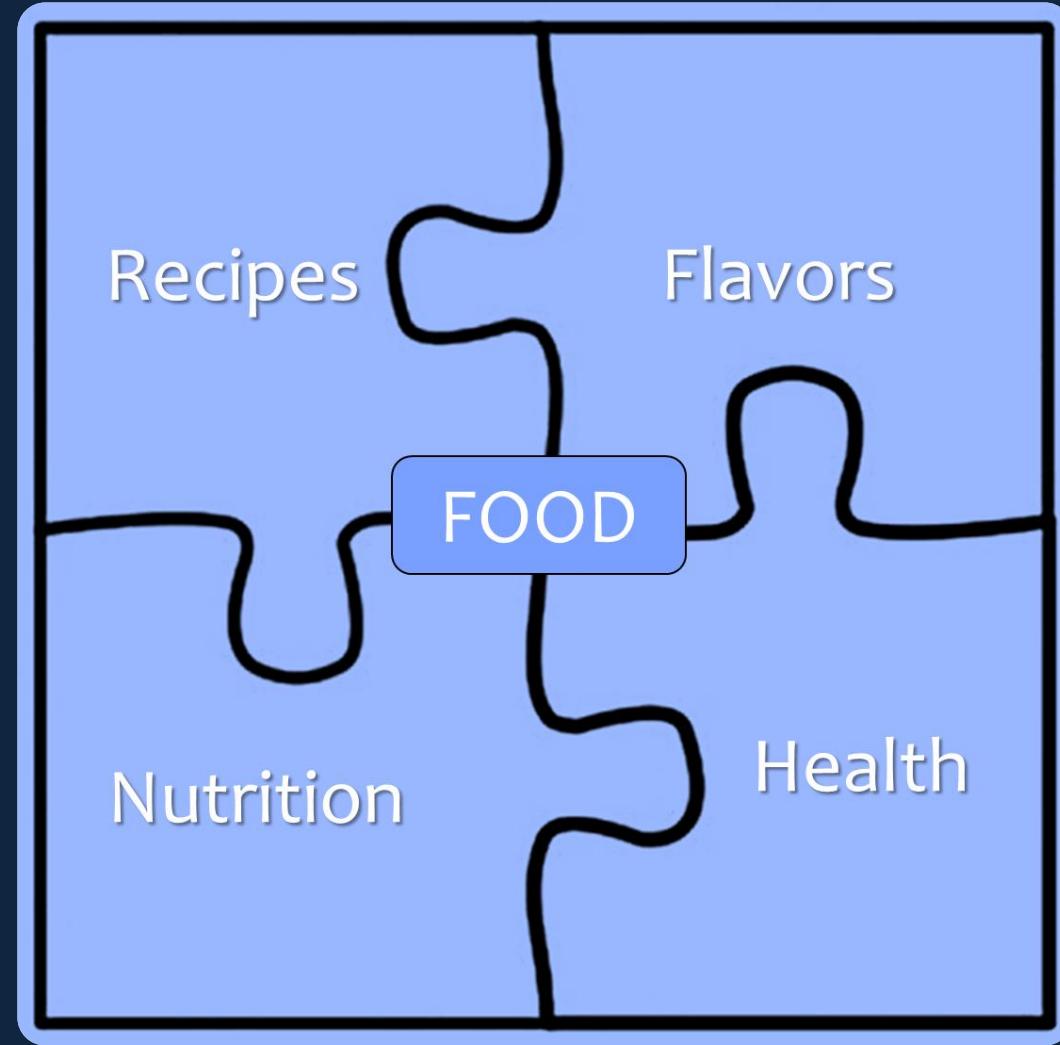


DATA

Making Food Computable



MAKING FOOD COMPUTABLE



TEDxIISERPune
x = independently organized TED event



Making Food Computable

TEDxIISERPune

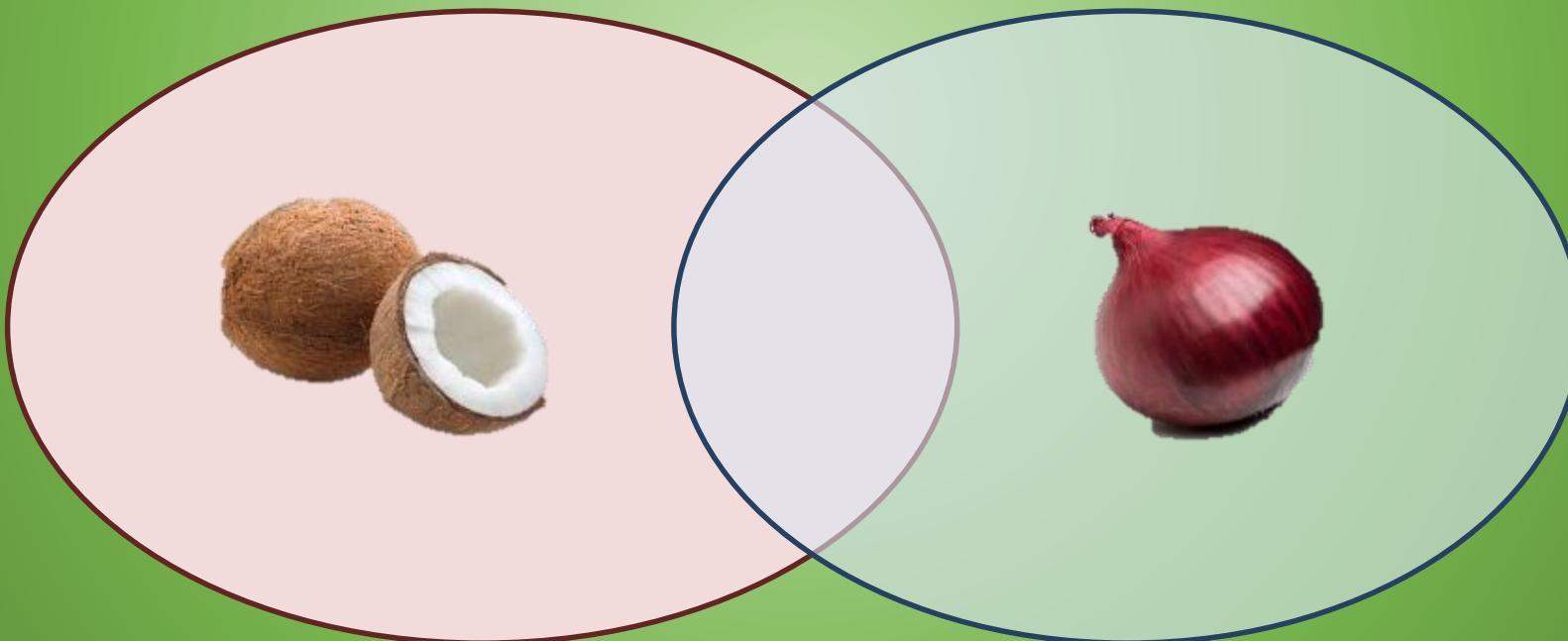


Why do we eat what we eat?

Why do we combine ingredients
in our recipes the way we do?

Food Pairing Hypothesis

Ingredients that **taste similar** tend to be **used together**
in traditional recipes



Ahn *et. al*, "Flavor network and the principles of food pairing", Scientific Reports (2011).
A Jain, NK Rakhi, G Bagler*, "Spices form the basis of food pairing in Indian cuisine", arXiv:1502.03815 (2015).

India—A potpourri of cultures with a rich and diverse culinary heritage

- ❖ Diverse
- ❖ Culture-rich
- ❖ Wellbeing-centric



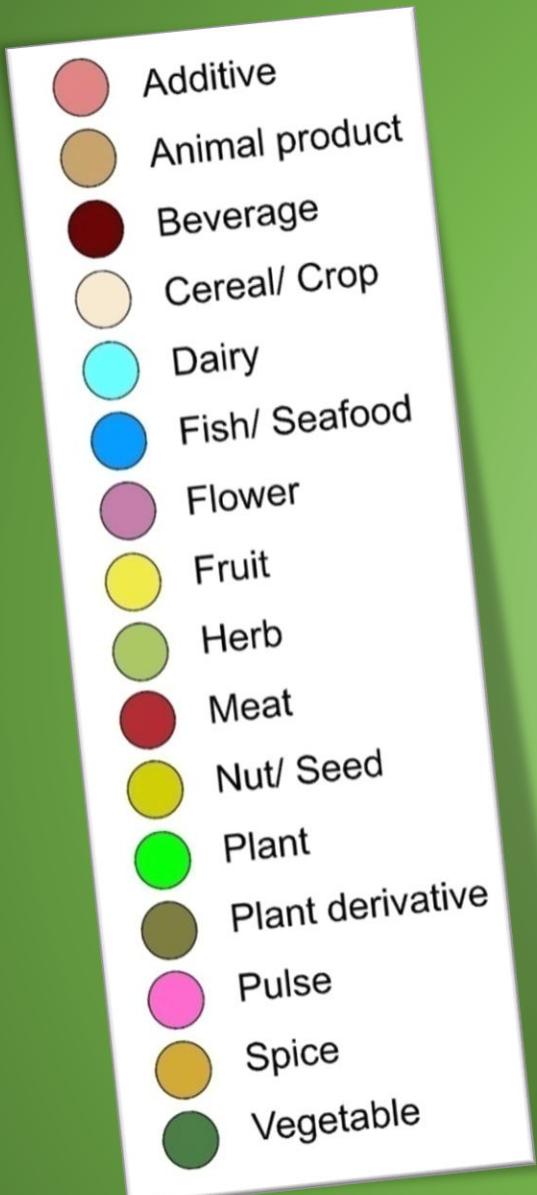


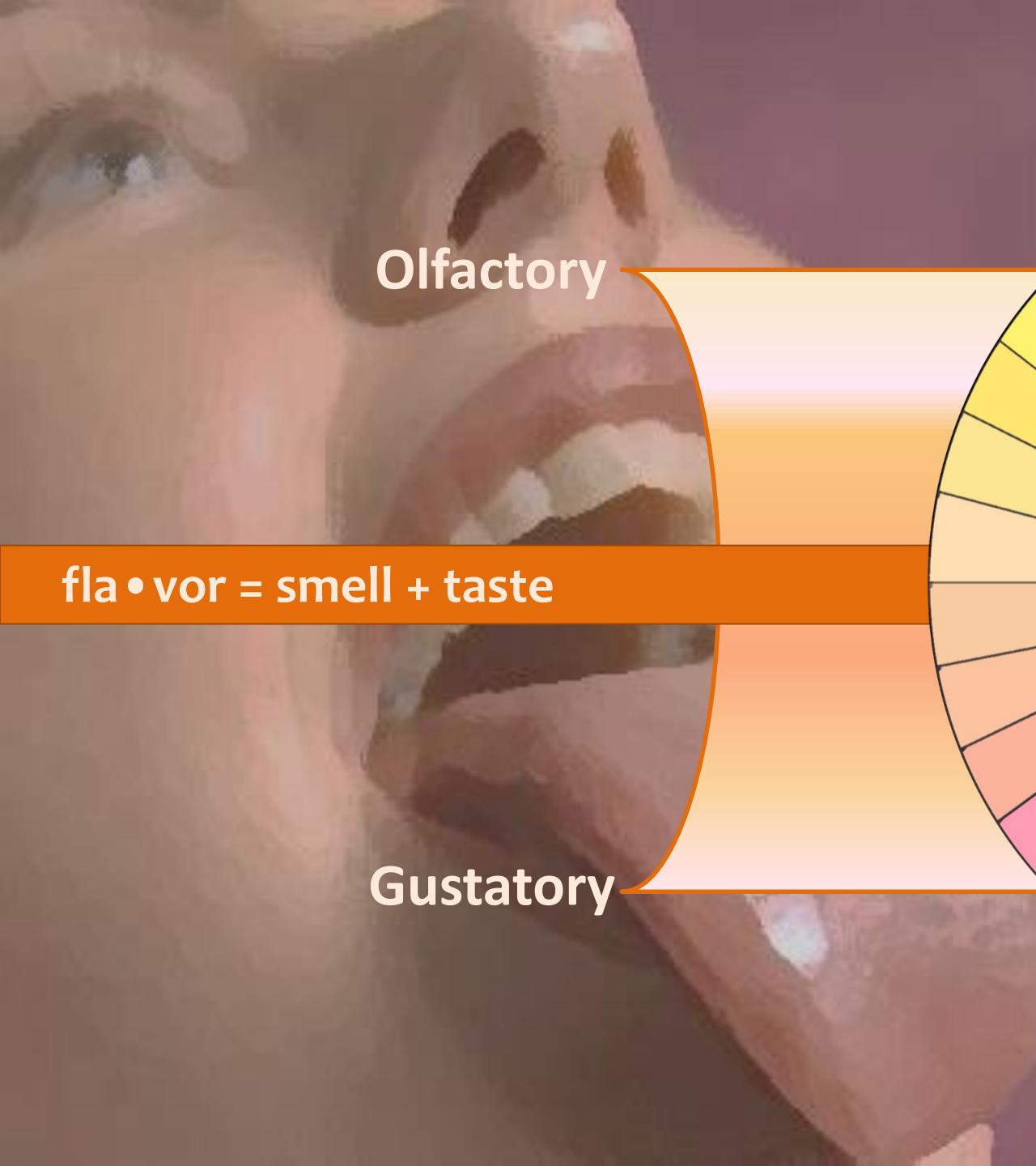
Data is the New Oil

Recipes & Ingredients

2543 Traditional Indian Recipes (TarlaDalal)

Regional cuisines: Bengali, Gujarati, Jain, Maharashtrian,
Mughlai, Punjabi, Rajasthani, South Indian.

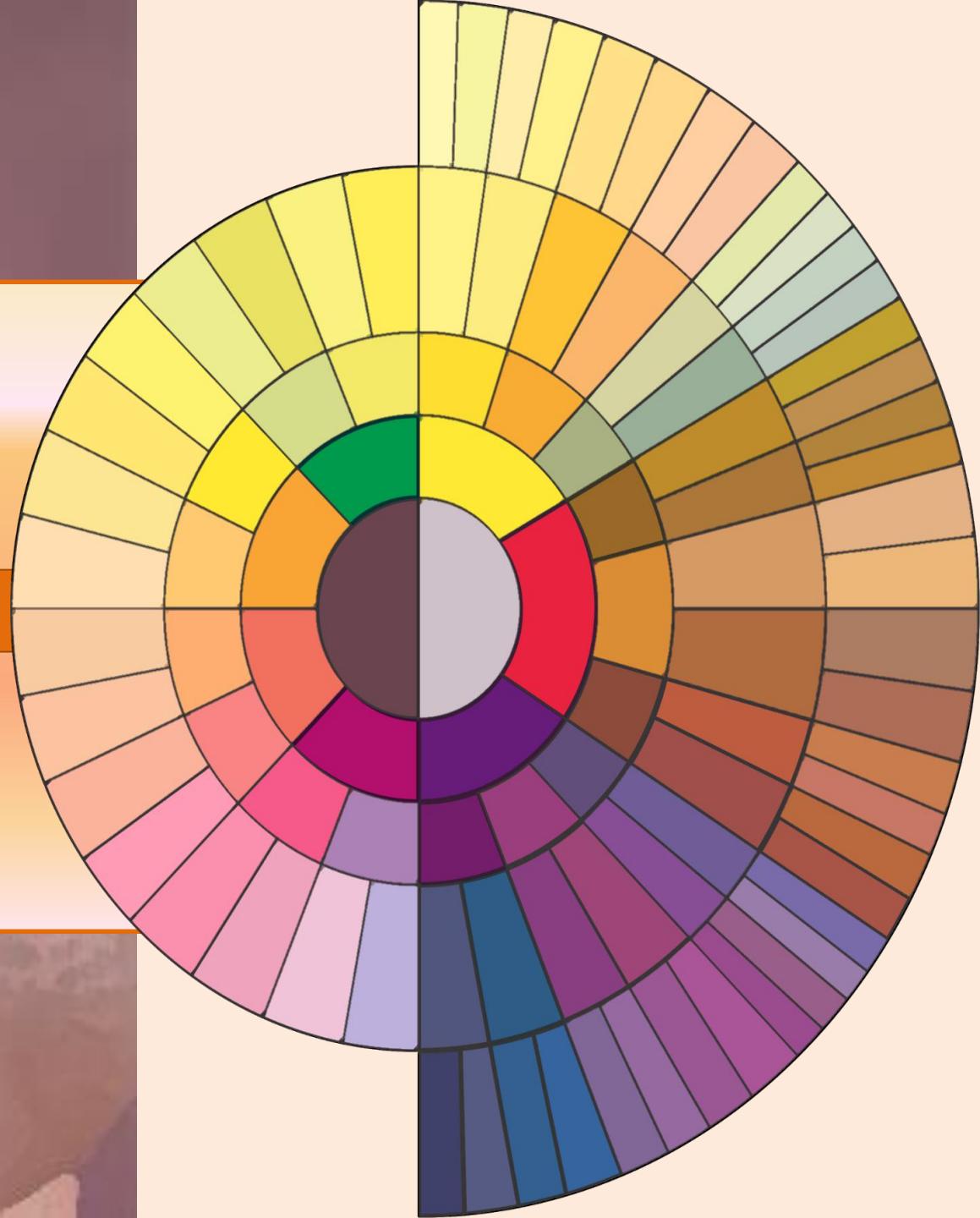


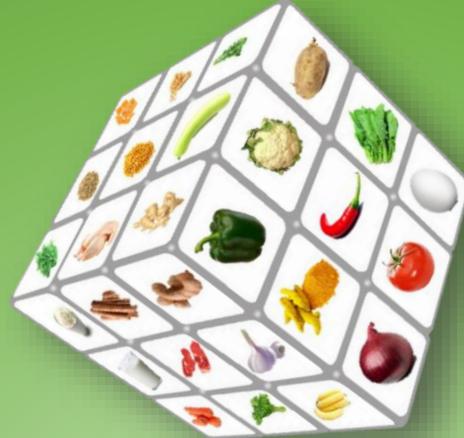


Olfactory

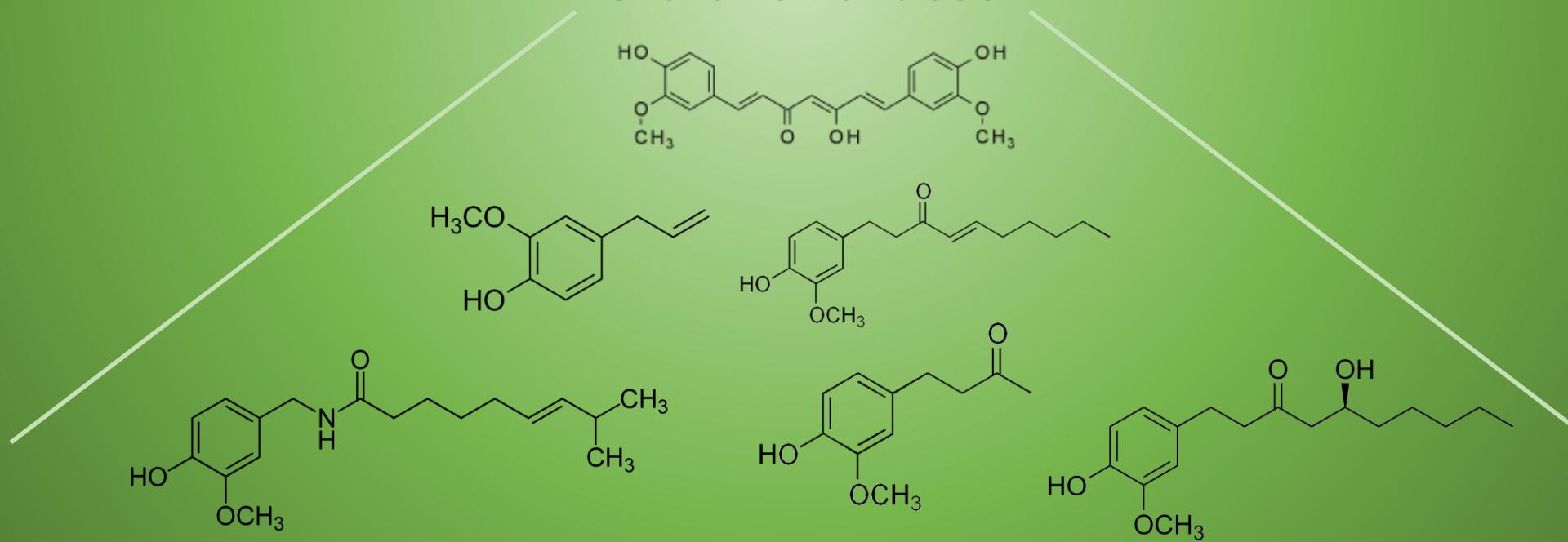
fla • vor = smell + taste

Gustatory





PubMed
PubChem
Fenaroli's Handbook



Food Pairing

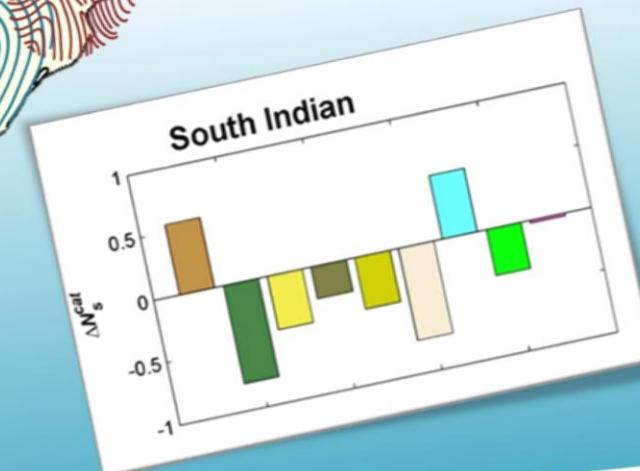
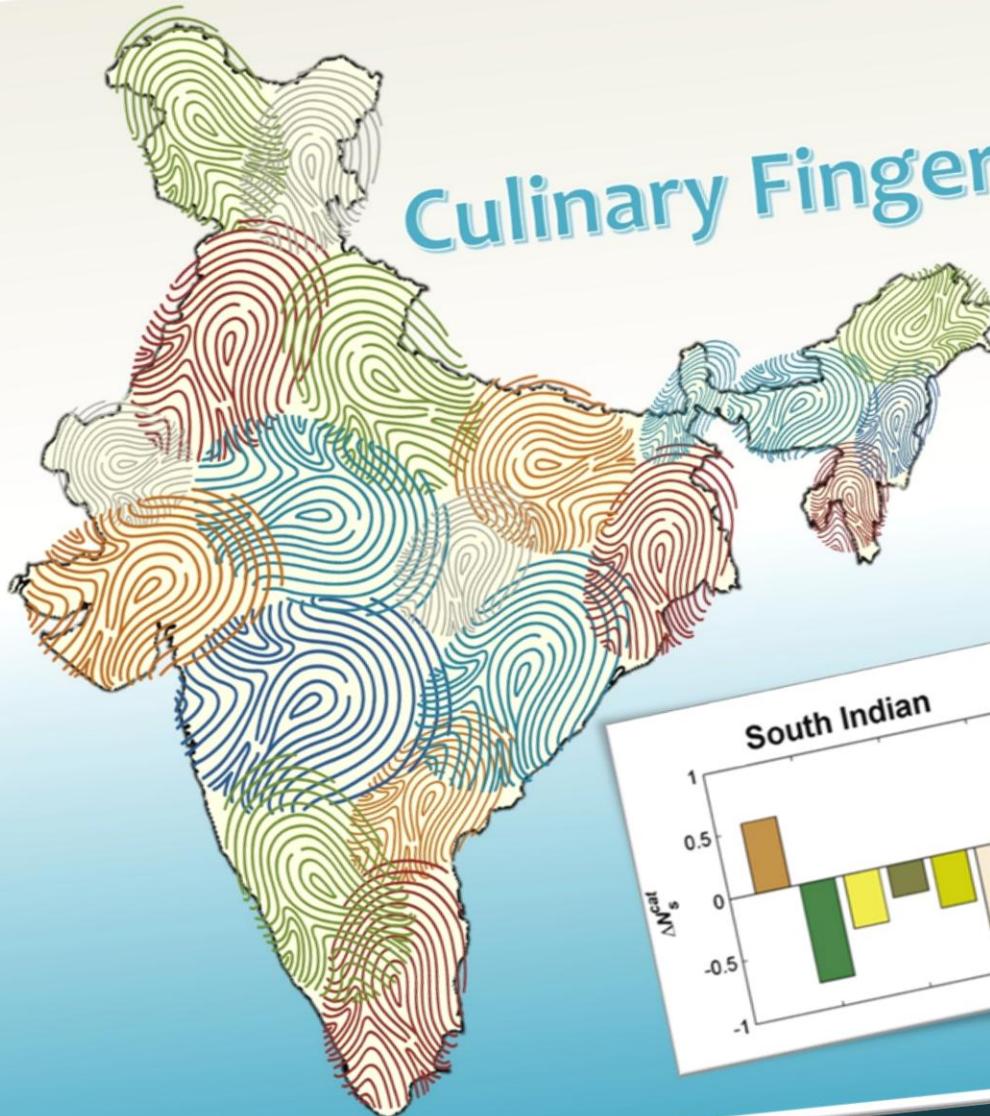




Spice

The Taste of India

Culinary Fingerprints



A View from Emerging Technology from the arXiv

Best of 2015: Data Mining Indian Recipes Reveals New Food Pairing

By studying the network of links between Indian recipes, computer scientists have discovered that the presence of certain spices makes a meal much less likely to contain ingredients with flavors in common.

MIT
Technology
Review

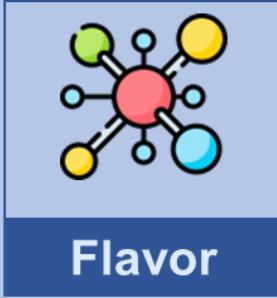
Best of 2015
MIT Technology Review







PLEASURE



PUBLIC HEALTH



COMMUNITY



CULTURE



SUSTAINABILITY



CIVILIZATION



Making Food Computable



RecipeDB

A structured repository of Recipes from across the globe



FlavorDB

A repository of flavor molecules found in food ingredients



SpiceRx

A platform for exploring the health impacts of culinary herbs and spices



DietRx

A platform for exploring the health impacts of dietary ingredients



SustainableFoodDB

A repository of carbon footprints of recipes



Ratatouille

Generate novel recipes with Artificial Intelligence



BitterSweet

A webserver for predicting BitterSweet taste molecules



SweetPred

State of the art ML algorithms to predict sweetness



WhatDish

A real-time dish detection app for diet management

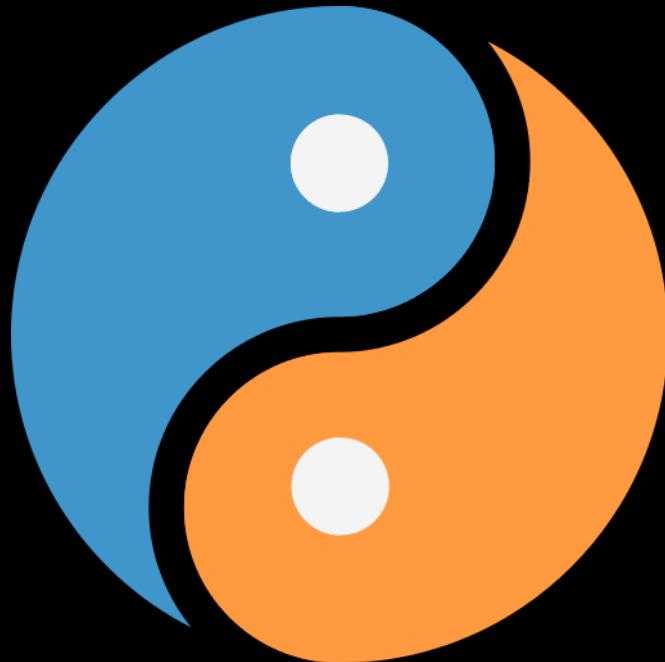


Ayurveda Informatics

A computable database of Ayurveda

Complex Systems Laboratory, IIT-Delhi

The Ground Zero of Computational Gastronomy



Making Food Computable

<https://cosylab.iiitd.edu.in>

Culinary Fingerprints of the World Cuisines





The food business is using ingredients' data and algorithms to understand how to create unique recipes, how to ensure sustainable food

CONSUMING PASSION: Data and delicious food

USING DATA TO PREDICT THE NEXT BIG FOOD ITEM

 Ingredients in an Indian recipe are different from an American recipe. Capturing this difference by digging through the data of recipes, ingredients, and flavour molecules and showing it in a statistical pattern was unusual till we did it. Our research became popular because nobody had looked at the uniqueness of Indian recipes in a quantifiable manner.



Ganesh Bagler |
PROFESSOR,
INDRAPISTHA
INSTITUTE OF
INFORMATION
TECHNOLOGY
DELHI



 Predicting the next big food item or food group is a big part of using data. We look at items that people are posting, blogging, and new flavours that are being launched, to predict the next thing. This year, people are high on millets. Lentils, fermented food and plant-based proteins from lentils and mushrooms will be big after millets.

Akshay Malhotra |
CHEF ENTREPRENEUR,
AND PARTNER, TAGTASTE

 I use (Ganesh) Bagler's work (in computational gastronomy) to combine foreign ingredients for Indian dishes. A miso marinated chicken tikka is awesome and it was possible because of the common flavours of miso with Indian spices

Parvinder Singh Bali | CORPORATE CHEF, OBEROI CENTRE OF LEARNING AND DEVELOPMENT



DECODING FLAVOURS

► A pioneer of computational gastronomy in India, Dr Ganesh Bagler is helping chefs reinvent dishes, writes Tania Bhattacharya

If you are a foodie and love experimenting with flavours, how does a shrimp and egg salad sound for lunch? Or a tenderloin steak, just out of the oven, topped with a dollop of cheese, melted with Camembert, to be precise. And did you know mushrooms go smashingly well with nearly every common fruit, from apples to apricots and even coconuts?

These bold pairings may sound unusual or at least unusual to most people, but Dr Ganesh Bagler, a professor at the Indraprastha Institute of Information Technology, New Delhi, thinks otherwise – and he has the data and research to back it up. He is the computational gastronomy expert who has taken his food and drink know-how from his ground-breaking work on flavour molecules and their corresponding databases, FlavourDB. His laboratory has also developed DietDB, an archive of nearly 2,000 foods, their chemical and genetic compositions, and their effect in health, which can enable drug and disease interventions. (Ayurvedic diets are a prime example of the belief in healing via appropriate foods.)

"The power of food and good taste is magic," says Bagler. Already, chefs such as Arun Arora of Restaurant Arava, in Bangkok, are using Bagler's research to fuel their own food experiments. "What I find amazing about Bagler's research is that this approach actually enables us to know exactly what makes up a cuisine – the things that make up a cuisine Indian," Arora, who is the first Indian woman with a Michelin star to her name, tells *The National*.

Bagler, who is now considered the pioneer of computational gastronomy in India, credits his journey for his success. Having studied various subjects from graduate to postdoctoral studies – quantum mechanics, computer science, computational biology, computational



Computational gastronomy expert Ganesh Bagler. Photo: Deewakar Bagler, Tejaswi Hingorani, Food Forward India

flavours and ingredients we don't need."

Bagler's work is also critical to Arora's Food Forward India, a non-profit initiative that aims to bring the narrative around Indian food. "It fits into the framework by being a forward-thinking initiative, one that serves the purpose of educating a culture, and identifying, quantitatively, its identity," says project manager Manjula Girela.

Bagler, who is now considered the pioneer of computational gastronomy in India, credits his journey for his success. Having studied various subjects from graduate to postdoctoral studies – quantum mechanics, computer science, computational

biology, and blue cheese taste great and blue cheese taste great together because they share 73 flavours." Blumenthal's interest was piqued when he paired white chocolate and caviar, and hit the right notes. This led to comparing such as Foodpairing, with present thousands of combinations of ingredients for chefs to experiment with.

"Historically speaking, dishes have evolved over millennia from single-ingredient meals to complex ones," says Bagler. "Cooking techniques and creation of new dishes aside, why are some ingredients used together and others not?" This was one of the critical questions that led Bagler to expand his research. "Food science has been around, but it explored

food or how to enhance sensory enjoyment. Now, people are looking at food from a data perspective."

What Bagler did differently was focus on Indian food, which he found is different from other cultures because of the spices. Breaking down a collection of the late, celebrated Indian chef Tarla Dalal's recipes, Bagler realised that spices form the basis of food-pairing in Indian cuisine. Having divided various foods into six categories – vegetables, dairy, lentils, meats, etc – he saw that missing up items across all other sections did not cause much of a shift in flavour, but when the spices were shuffled, the taste changed entirely. For example, you could replace spinach

panner and there would not be much change in the dish, but if you replaced turmeric with cinnamon, the very essence of the preparation alter. "Spices are the molecular hub of Indian food," says Bagler.

In 2012, a team of researchers sent this study to international science journals, which episode it to an open source. In 2014, it was picked up by *MIT Tech Review*. This changed Bagler's life. "I only understood the academic value of this work, not its futuristic value," he says. "It took me a year to understand that this had led to the creation of a new field of study, and over the past five years, I've been developing the foundations of this area."

From 2,543 of Dalal's reci-

pes, Bagler's database has expanded exponentially. Not only is the data free to access online, but the information is also provided in an easy-to-use format, from the scientific names of elements to a comprehensive flavour, colour, possible pairings and health benefits. While Bagler consults for institutions such as the Indian Institute of Hotel Management and Symbiosis School of Culinary Arts, as well as a range of restaurants, chefs also have "significantly reduced time spent on developing new dishes," says Alka Malhotra, a chef, food consultant and former student of the Culinary Institute of America.

FlavourDB complements

chefs' instincts about pairing ingredients. And chef-cum-chef Manjula Gill and Akshayraj Jodha, director of Bagler's dessert, are successfully quantifying the knowledge that, until now, was only intuitively available to a cook – and everyone from chefs and diners to scientists



Garima Arora, above, of Restaurant Gar, is using Bagler's research to experiment with flavours, top. Centre, her strawberry-curd and hor wok (Thai lemon verbena) dish

intelligence will influence the food industry," he says.

This aspect is also key to Bagler's future experiments. "Can we encode the intelligence of a chef into a computer so that a computer feels a chef into thinking a recipe is real?" says Malhotra.

Can human creativity, which is at the heart of cooking, be reproduced using AI? It remains to be seen. For now, Malhotra's observations pertain to FlavourDB complementing chefs' instincts about pairing ingredients. And chef-cum-chef Manjula Gill and Akshayraj Jodha, director of Bagler's dessert, are successfully quantifying the knowledge that, until now, was only intuitively available to a cook – and everyone from chefs and diners to scientists



AI and Food Innovations | Paris

The cover of the Fall 2023 issue of Food Innovation Quarterly. The title 'Food Innovation Quarterly' is in large blue letters at the top, with 'FALL 2023' in green below it. To the right, there are four sections with titles: 'The Design of New Food Products: Generative AI & Behavioral bricks in action', 'AI driven Foodscapes: Sketching the Ecosystems of Food', 'Food, AI and Habitus : Capturing the Consumer Behavioral Surplus', and 'Food Innovation Report Rankings & Analytics p. 41'. Below the titles are four product images: a container of 'Critm Cekoo', a container of 'Ceardna Kit', a box of 'Criareet', and a container of 'Creat'. The bottom banner features the text 'AI and Food Innovation' in large pink letters, followed by 'A New Data Science of Food'.



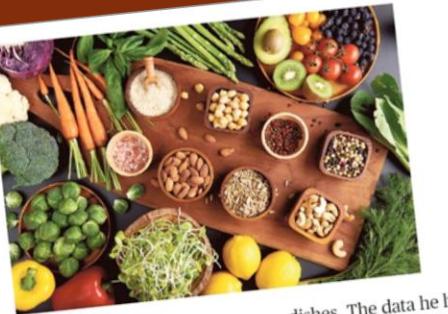
LSTC Golden Jubilee Celebrations



Computational Gastronomy *The Future of Food*

Ganesh Bagler

New culinary experiences
The technique helps in coming up with innovative flavour pairings.
(Below) Ganesh Bagler. SPECIAL ARRANGEMENT



Data-driven delicacies

Ganesh Bagler of Indraprastha Institute of Information Technology in Delhi aims to revolutionise our approach to food with data-driven Computational Gastronomy

Ananya Desikan
ananya.a@thehindu.co.in

Ganesh Bagler from Indraprastha Institute of Information Technology, Delhi has a new lens to look at gastronomy, which he believes can change the way we approach food. “Why do we eat what we

eat? or Why do we pair ingredients the way we do? These are the questions we started with to get to where we are now. Computational Gastronomy is a data-driven way of looking at food to create new culinary experiences. We get into the ingredients at a molecular level to find answers,” he says.

Along with his team of

researchers and developers, Ganesh has put together 10 repositories and applications based on this approach. Data has been compiled from one lakh eighteen thousand recipes from 26 world regions and 74 countries.

Understanding food

This methodology is used to study the evolution of cuisines and is also believed to be helpful in coming up with innovative flavour pairings. “When we try to understand how contemporary Indian food came into being, it is similar to learning about the evolution of language. We place the data available to us in a multi-dimensional space and study similarities and dissimilarities with other global cuisines,” he claims.

While this methodology helps understand the evolution of Indian cuisine, chefs who use this approach in their business believe that this also contributes to the evolution.

Garima Arora, the first female-Indian chef to win a Michelin star for her restaurant Gaa in Bangkok, says Bagler scientifically explains Indian food, adding, “This is immensely helpful to chefs chasing new flavours and

dishes. The data he has collected helps in coming up with new ingredient combinations that work. This can revolutionise Indian food.”

In the FlavourDB repository, there are 25,595 flavour molecules that contribute to the taste or odour of an ingredient. By browsing through the extensive flavour network graphic or visual search option, one can look for flavour pairings based on the molecules the ingredients share. For instance, fenugreek shares more than 100 flavour molecules with ingredients like tea, soybean, papaya, guava and even cocoa.

“Our SustainableFoodDB repository gives the carbon footprint of each recipe, ingredient and also provides alternative ingredients,” he says.

While going through the enormous amounts of data can be overwhelming, Ganesh informs us that there are applications like Ratatouille that can generate recipes based on the ingredients available. Ganesh says that

when he deconstructs cuisines into recipes and ingredients to molecules, he feels like Dennis The Menace who pulls his toys apart. But he believes, “It is not just the data but algorithms that we have that can make a difference in the future of food.”



CHEMISTRY WORLD



Currying flavour

“The intersection of food science, data and computation is one of the most exciting frontiers, and your work is right in this sweet spot.”

Sam Arbesman
Scientist in Residence, Lux Capital



FoodBytes API

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Digital Building Blocks of Food

Computational Gastronomy APIs

Explore APIs

Foodoscope Technologies Private Limited

Center for Sustainable Food



DIGITALLY DELICIOUS



The adventure of making food computable

GANESH BAGLER

India Science Book Fellowship 2024

Course Structure

This course introduces the emerging science of Computational Gastronomy, a blend of food, data science, and computational techniques.

The following are the specific objectives that the course aims to fulfill:

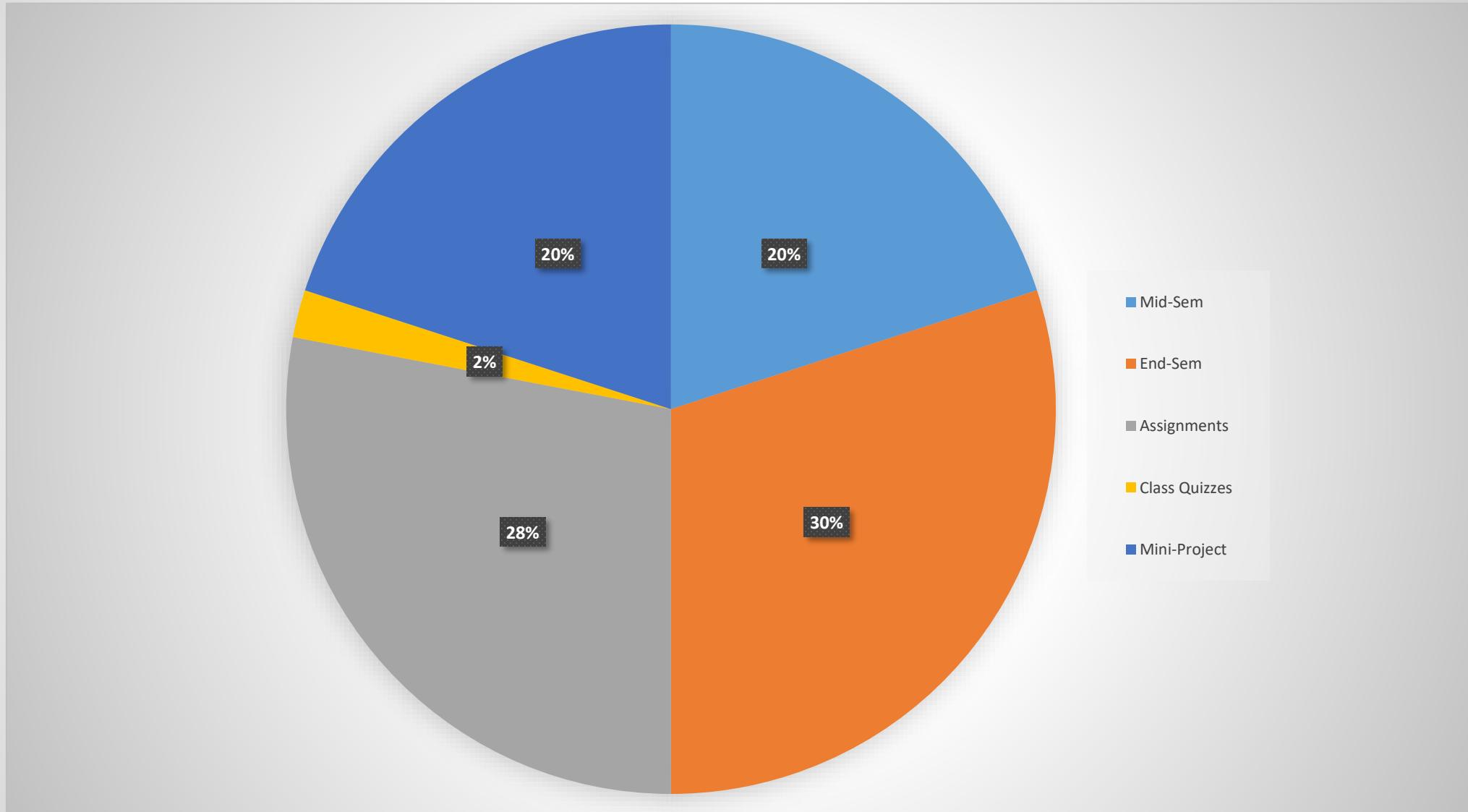
- (a) Introduction to **the science of food**;
- (b) Exposition to **culinary data**;
- (c) **Analysis and visualization** of culinary data (statistical analysis, text mining, natural language processing, machine learning);
- (d) Exposure to **challenges and opportunities**.

Post Conditions

- Students are able to **explain** the basic tenets of the science of food and the scope for data-driven analysis.
- Students are able to **explain, analyze, and interpret** major aspects of **culinary data**.
- Students are able to **design classification and prediction models** from culinary data.

Desirable Pre-requisites: A good command of programming (Python) and data analysis. Exposure to statistical analysis, data visualization, machine learning, and natural language processing.

Percentage Division of Evaluation Criteria



FAQs

- **What are class quizzes?**
→ Simple MCQs through Google Forms, based on the content being discussed in the class. **Weightage: 2%**
- **What will be the nature and logistics of mini-projects?**
→ **Teams of 1-3 students working on projects for at least 3 weeks** touching on aspects of data scraping & organization, data analysis, visualization, ML models (classification, regression), design of Android apps, web apps, technology leverage (blockchain, NFTs, NLP, NER, Ontology, Computer Vision, etc.), Kaggle challenge design, external API utilization, browser plugins, game design, library design, instrumentation/Arduino, social science questions, design challenges, exploratory research, and entrepreneurship.
Evaluation: Based on project-specific deliverables. **Weightage: 20%**
Opportunities: Hackathon, project-and-poster presentations in the Symposium, interactions with potential employers from industry, government & policymakers.
- **What will be the structure of the mid-sem and end-sem exams?**
→ A combination of a small number of MCQs with partial negative marks (correct: +1, wrong: -¼) and short-answer questions. **Weightage: Mid-Sem (20%) & End-Sem (30%)**
- **What will be the nature and logistics of assignments?**
→ **Teams of 1-2 students working on three assignments (2 best of 3)** involving experiential work & coding-based questions involving data collection, analysis, visualization, and problem-solving. **Weightage: 28%**
- **Is attendance mandatory? Are there points for attendance?**
→ NO!!

Week	Lecture Topic	COs Met	Assignment/Exercise
1 & 2	Introduction to the science of food: History of food, Cultural influences, Impact of Technology, Role of data science. Cooking-- the craft and magic: Session by a practicing chef highlighting the need for scientific understanding of food. Role of Computational Gastronomy in achieving data-driven food innovations.	CO1	6 hours (Reading Material)
3 & 4	Culinary Data Analysis [Recipes]: Recipes data structure, Data compilation and curation, Visualization and analysis, The structure of world cuisines, Ingredient popularity and uniqueness.	CO2	5+2 hours (Reading Material + Exercises + Assignment-1)
5 & 6	Culinary Data Analysis [Flavors]: Flavors data structure, Data compilation and curation, Visualization and analysis, Molecular composition of ingredients, Sensory attributes of taste and odor.	CO2	5+2 hours (Reading Material + Exercises)
7 & 8	Culinary Data Analysis [Food Pairing]: Food pairing, Hypothesis testing, Culinary fingerprints, Comparative analysis of world cuisines, Data visualization.	CO2	5+2 hours (Reading Material + Exercises)
9	Culinary Data Analysis [Health-Nutrition]: Nutritional data analysis, Health impacts of food, Food-disease associations, Data analysis and visualization.	CO2	3+2 hours (Reading Material + Exercises + Assignment-2)
10 & 11	Modeling and Prediction [Cuisine]: Cuisine classification, Features selection, Implementation of machine learning models, Comparison with the state of the art, Identity of a cuisine, Transforming a recipe from one cuisine to another.	CO3	5+2 hours (Reading Material + Exercises)
12 & 13	Modeling and Prediction [Flavor and Fragrance]: Prediction of taste and odor; Features selection, Implementation of machine learning models, Comparison with the state of the art.	CO3	5+2 hours (Exercises + Mini Project)
14	Opportunities and Challenges: Challenges for building the foundations of a data-driven culinary science; Scope for compilation of culinary data from textbooks, scientific literature, traditional knowledge sources, web-based sources (Twitter, Culinary forums, Reviews, Restaurant menus, Food service aggregators, FMCG data repositories, etc.); Opportunities in food-tech industry; Entrepreneurship; Policy making; Food security; Public Health, and Sustainable Development.	CO1, CO2, CO3	3 hours + Mini Project Presentations

Lectures & Office Hours

When: Tuesdays and Fridays, 11am—12:30pm

Where: CO2 (Old Academic Block, Ground Floor)

Office Hours:

Tuesdays (9—10am & 12:30—1:30pm) &
Fridays (9—10am & 12:30—1:30pm).

A305 (R&D Block)

Computational Gastronomy



Infosys Centre for
Artificial Intelligence

GANESH BAGLER



A Data-driven Science of Food



Computational Gastronomy is a data science that blends food, data, and computation for data-driven food innovations.

Impact of Technology on Food



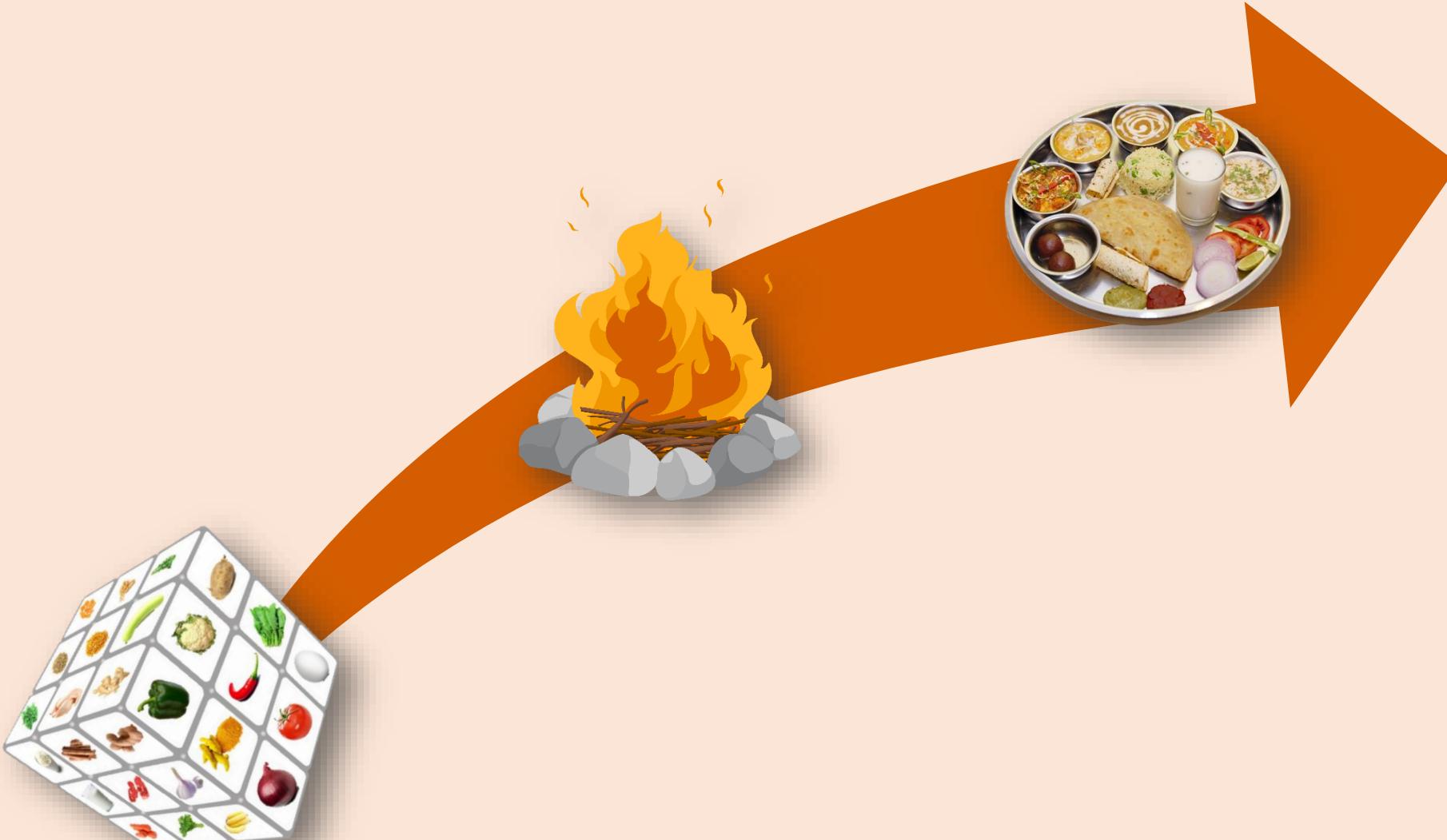
Scientific American, June 2015

Ganesh Bagler

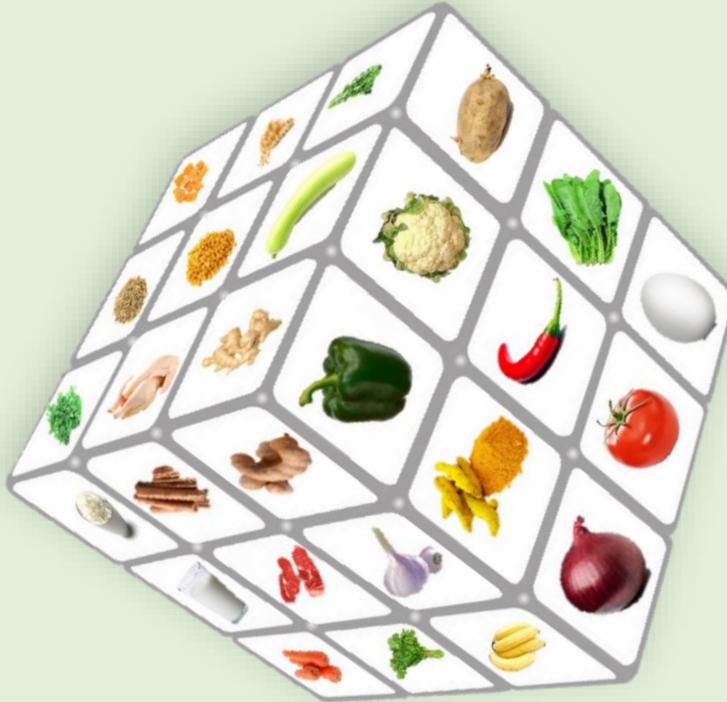
Center for Computational Biology, IIIT-Delhi, New Delhi.

- What is food?
 - What is technology?
-
- What are the ways in which food touches our lives?

Food is magic!

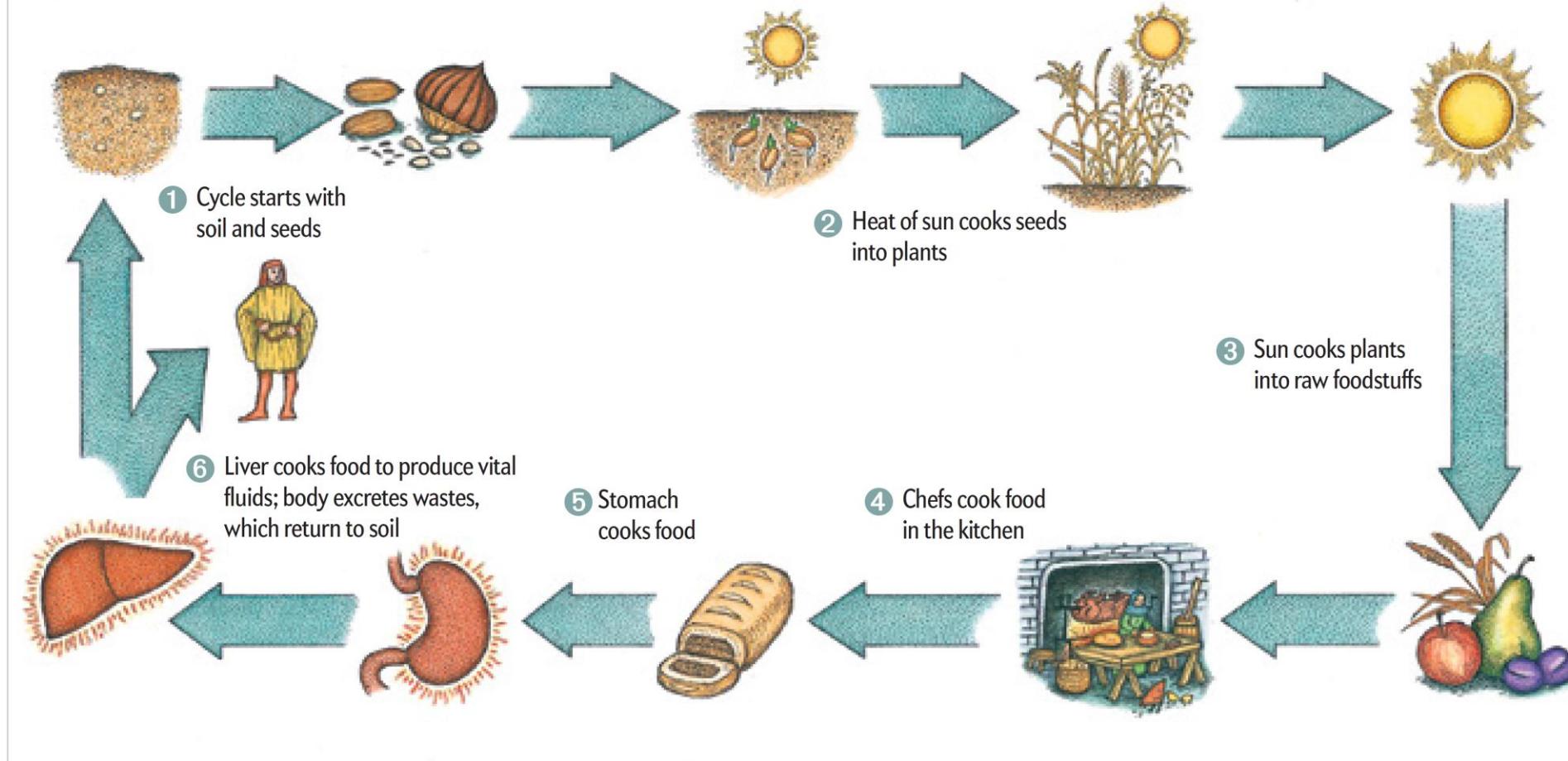


Food is Nutrition



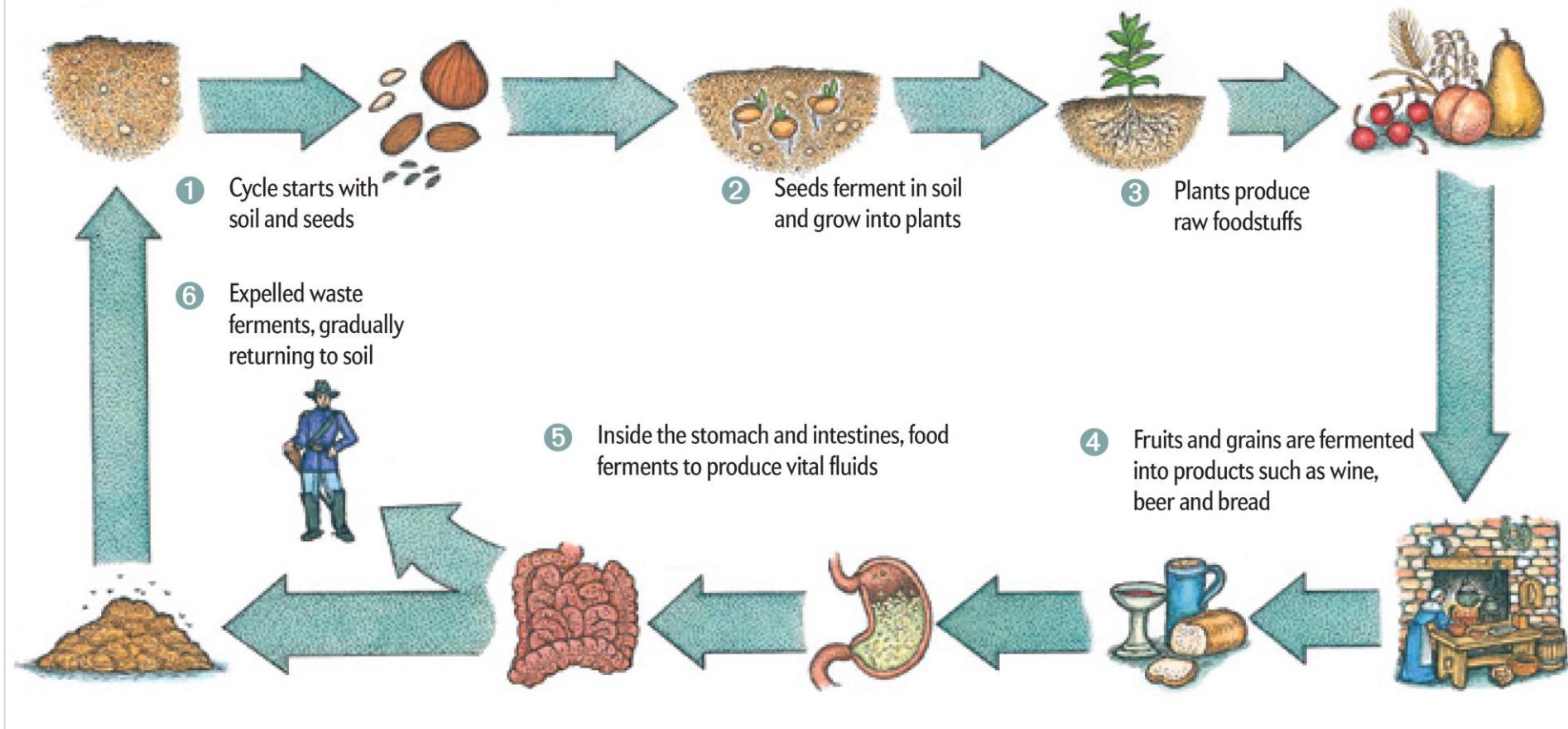
The Cosmic Culinary Cycle before 1650

... in which cooking was believed to be the central process of life.



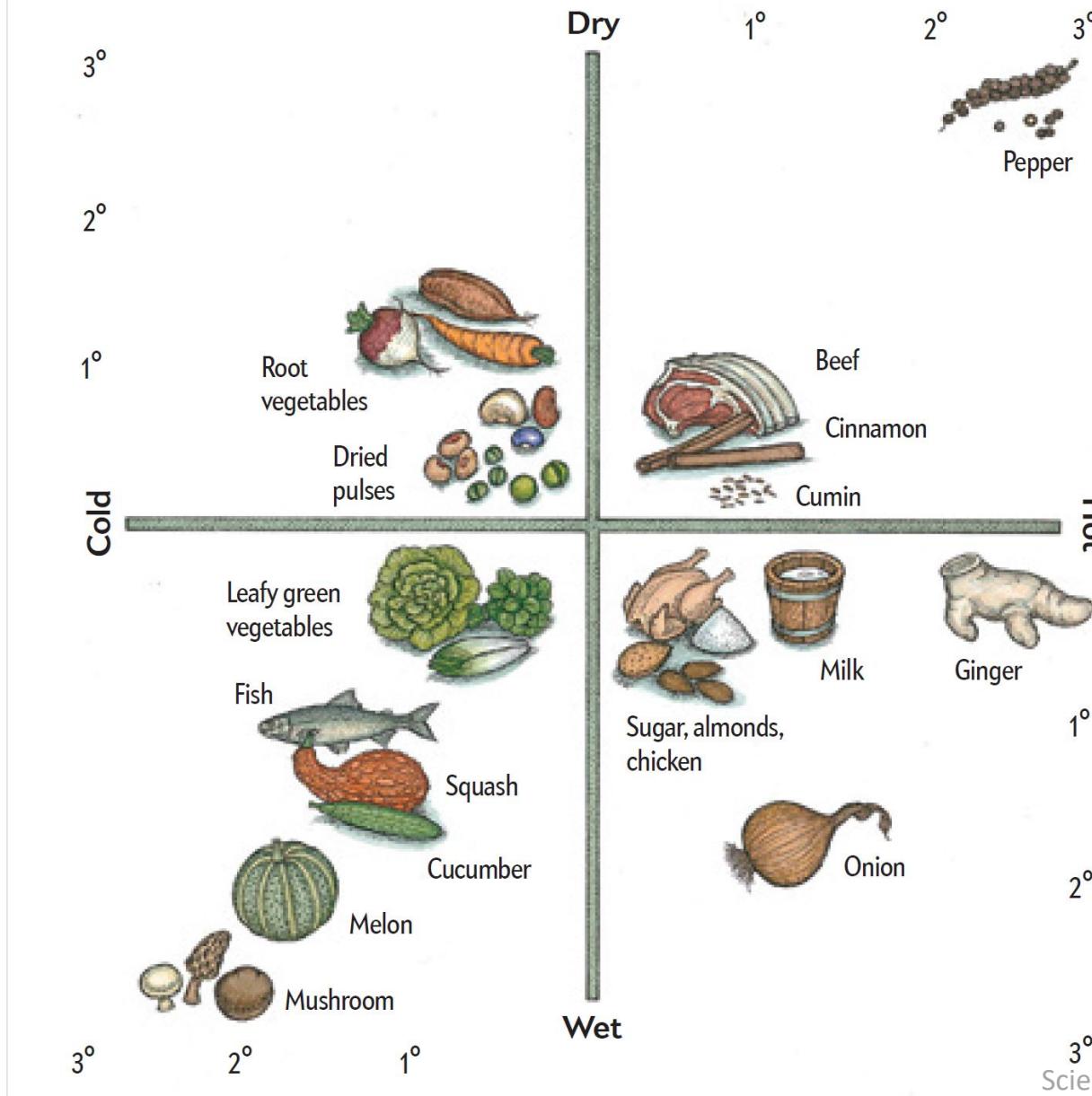
The Cosmic Culinary Cycle after 1650

... in which fermentation was believed to be the central process of life.



16th-Century Classification System

... in which foods were assigned degrees of heat, coldness, wetness and dryness.



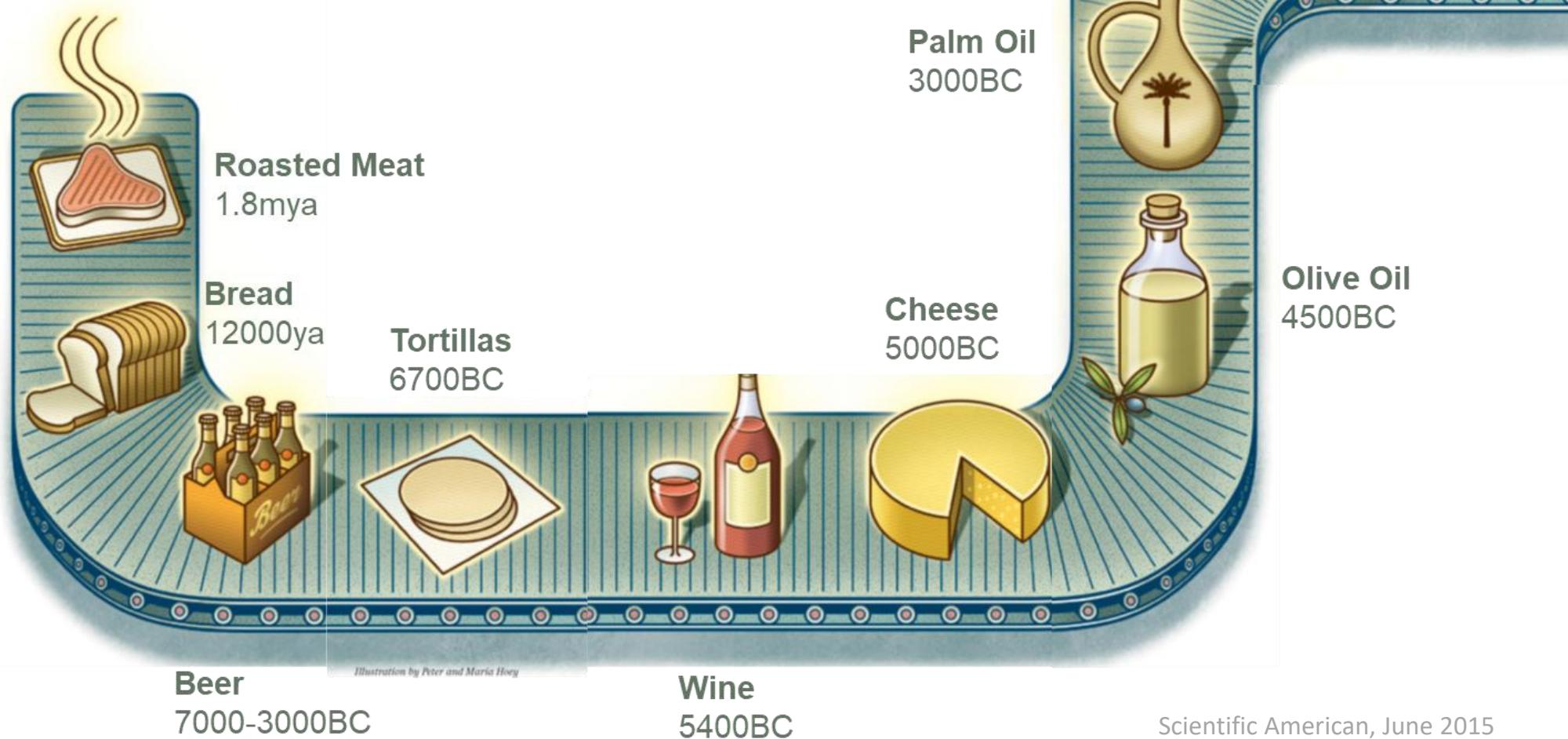
Processed Food



Processed Food

- ❖ It is the **dark force**, we're told, **behind the obesity epidemic** and the death of the family farm.
- ❖ But humans have been '**processing**' food **ever since we learned how to cook, preserve, ferment, freeze, dry or extract**.
- ❖ **Processed food** has powered the evolution of the species, the expansion of empires, the exploration of space.

Processed Food



Chocolate
1900BC



Bacon
1500BC

Jiang
1000BC



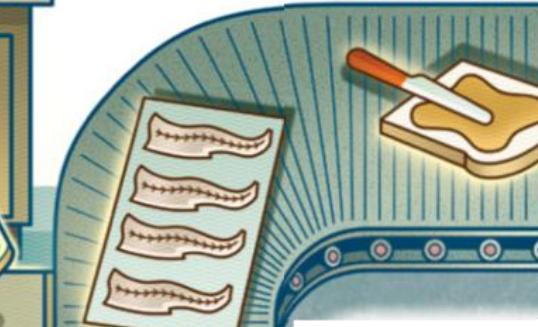
Sugar
500BC



Mustard
400AD



Kimchi
700AD

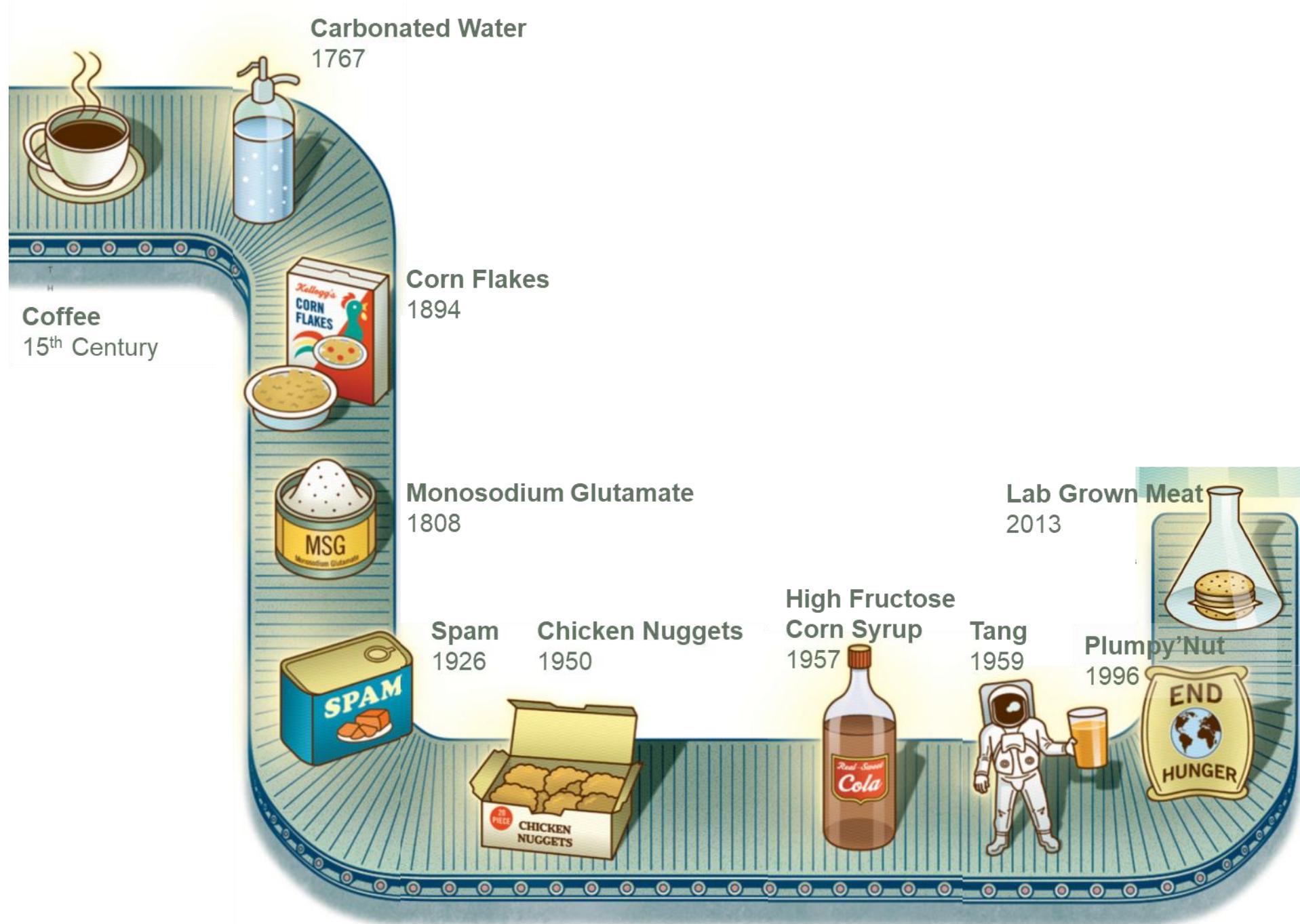


Peanut Butter
15th Century

Salt Cod
10th Century

Tofu
965AD

Sushi
700AD



Cooking Innovations: Refrigeration



Google Images

Cooking Innovations: Kerosene Stove



Cooking Innovations: LPG



Cooking Innovations: Microwave



Cooking Innovations: Pressure Cooker



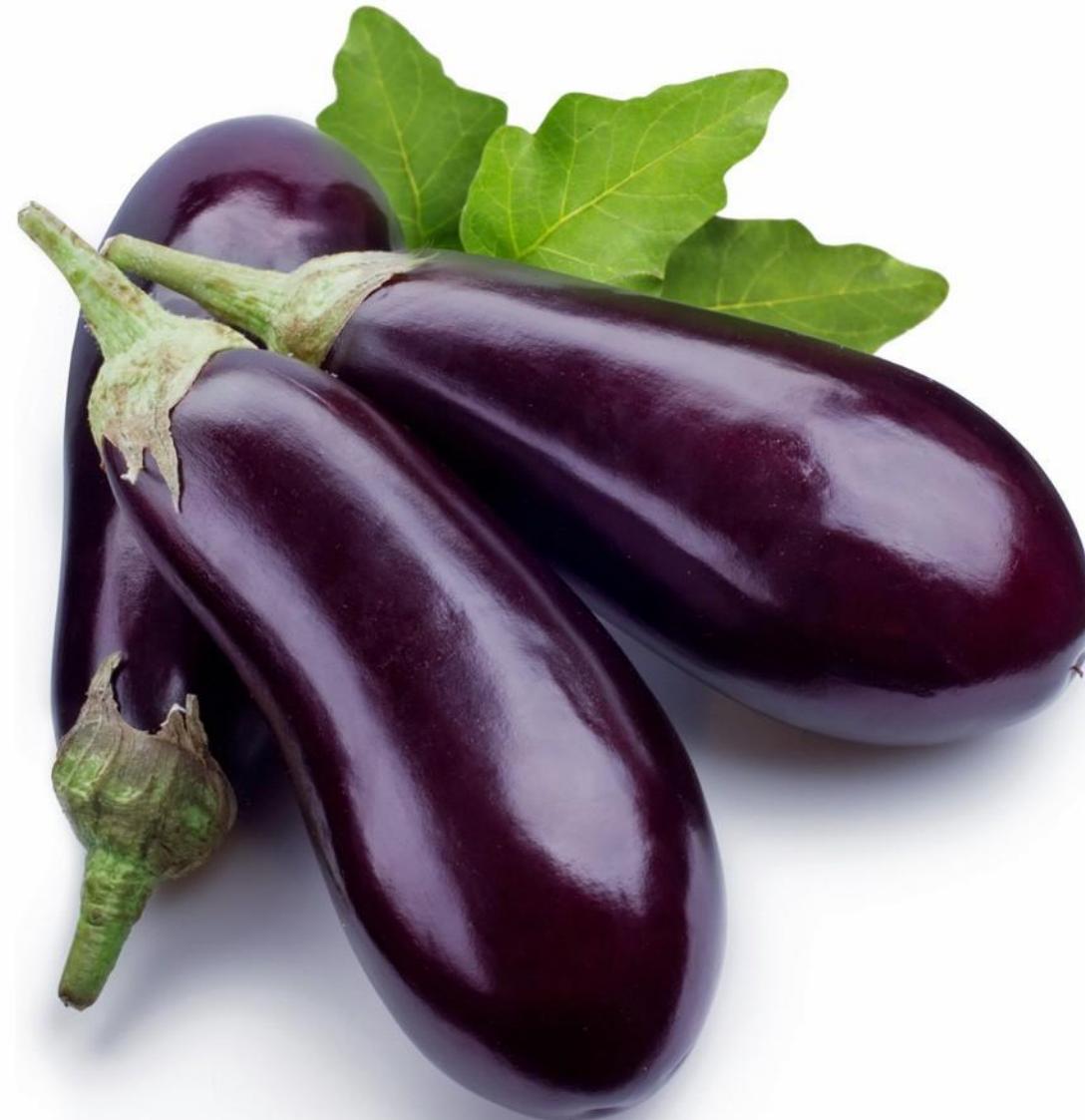
Cooking Innovations: Utensils



Sugar



Genetically Modified Organisms



The Impossible Burger



Cooked – Michael Pollan

A NEW YORK TIMES BESTSELLER

MICHAEL POLLAN

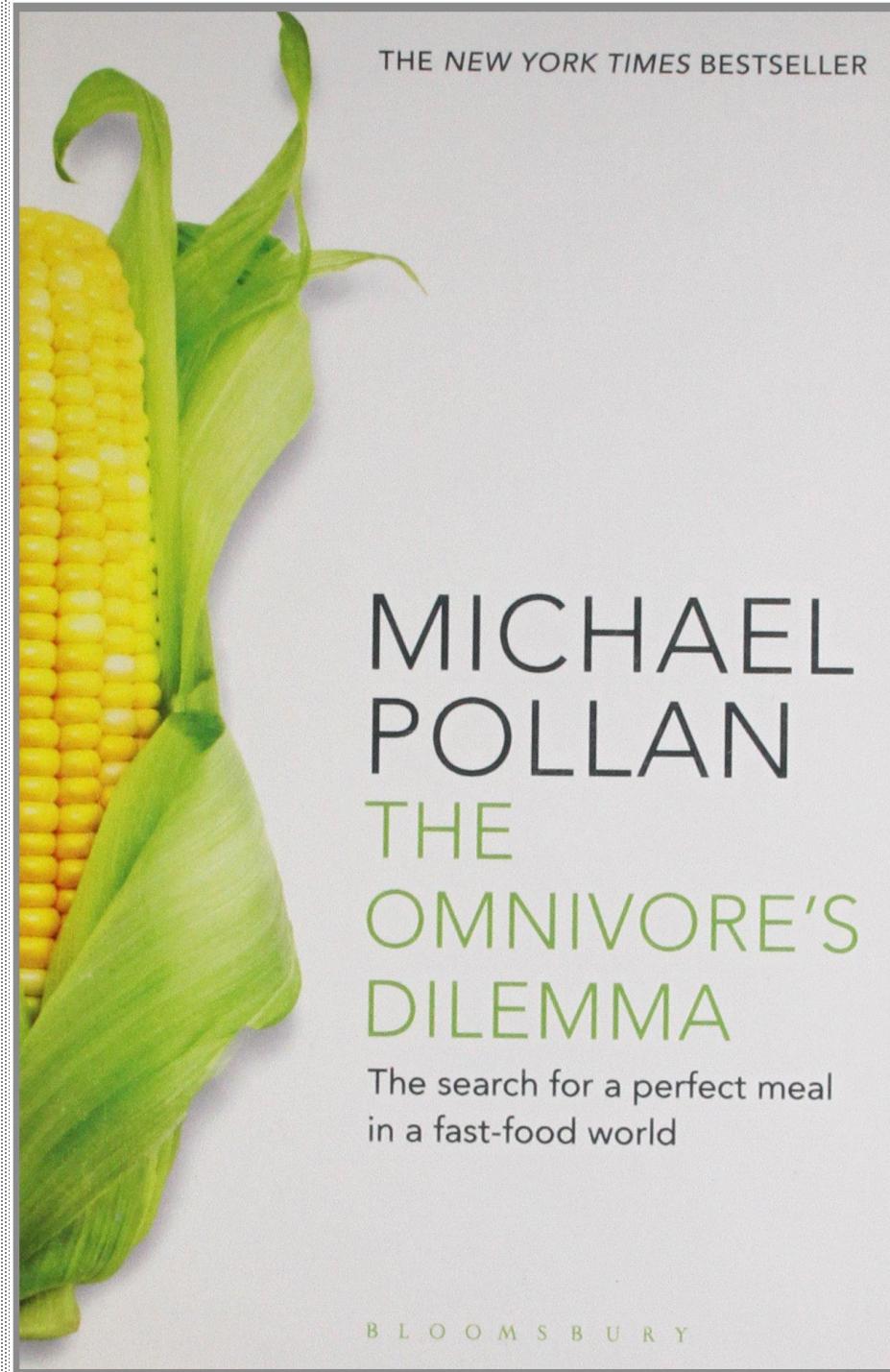
Author of THE OMNIVORE'S DILEMMA



COOKED

A NATURAL HISTORY *of* TRANSFORMATION

The Omnivore's Dilemma –
Michael Pollan



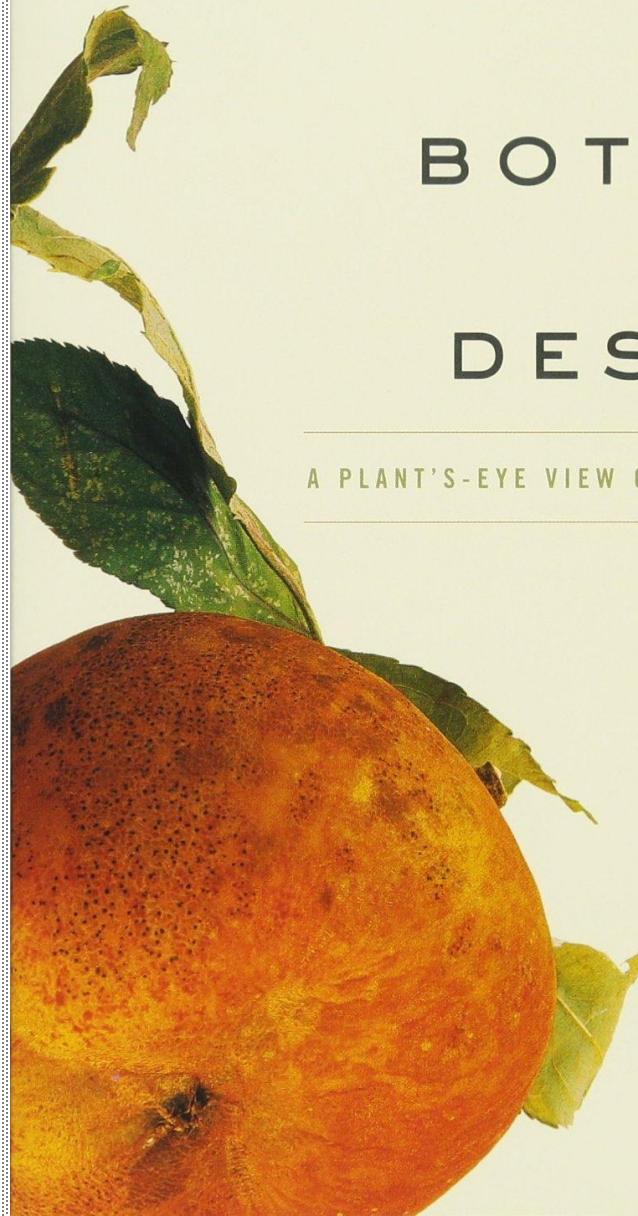
The Botany of Desire— Michael
Pollan

MICHAEL POLLAN

NEW YORK TIMES BESTSELLING AUTHOR OF *THE OMNIVORE'S DILEMMA*

THE
BOTANY
OF
DESIRE

A PLANT'S-EYE VIEW OF THE WORLD



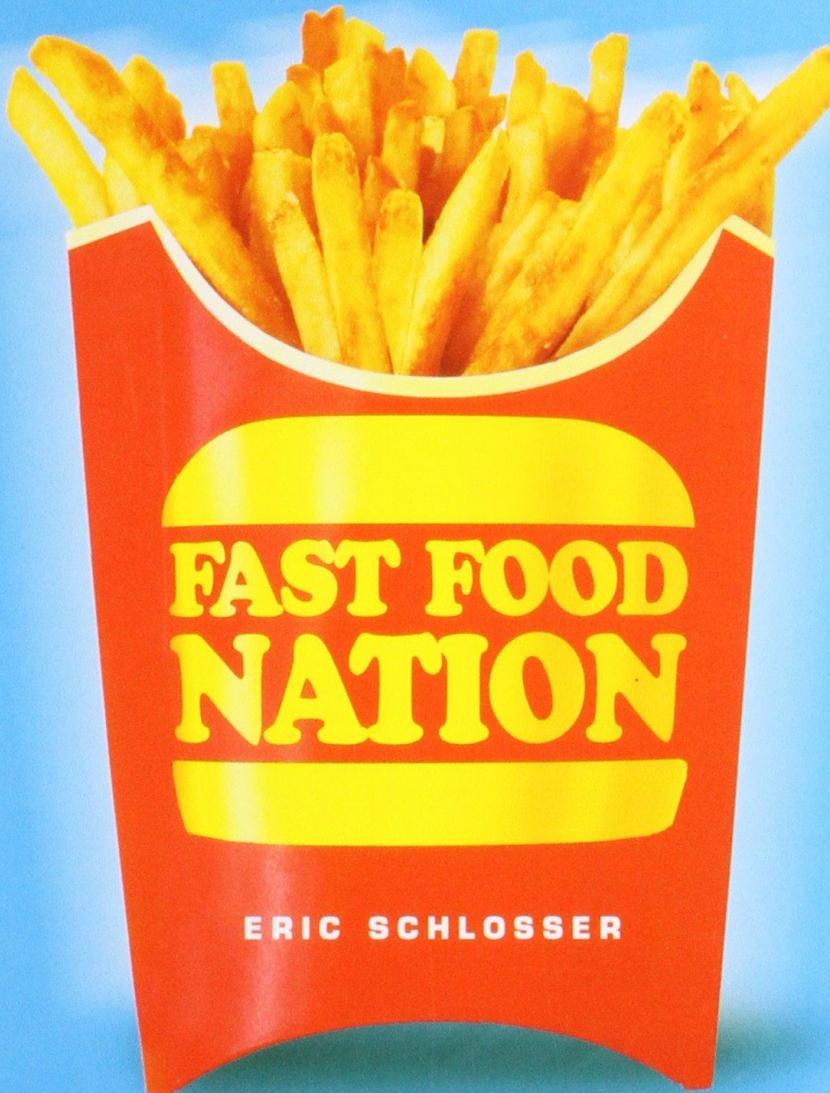
"Pollan shines a light
on our own nature as well
as on our implication in
the natural world."
—*The New York Times*

Fast Food Nation – Eric Schlosser

THE INTERNATIONAL BESTSELLER

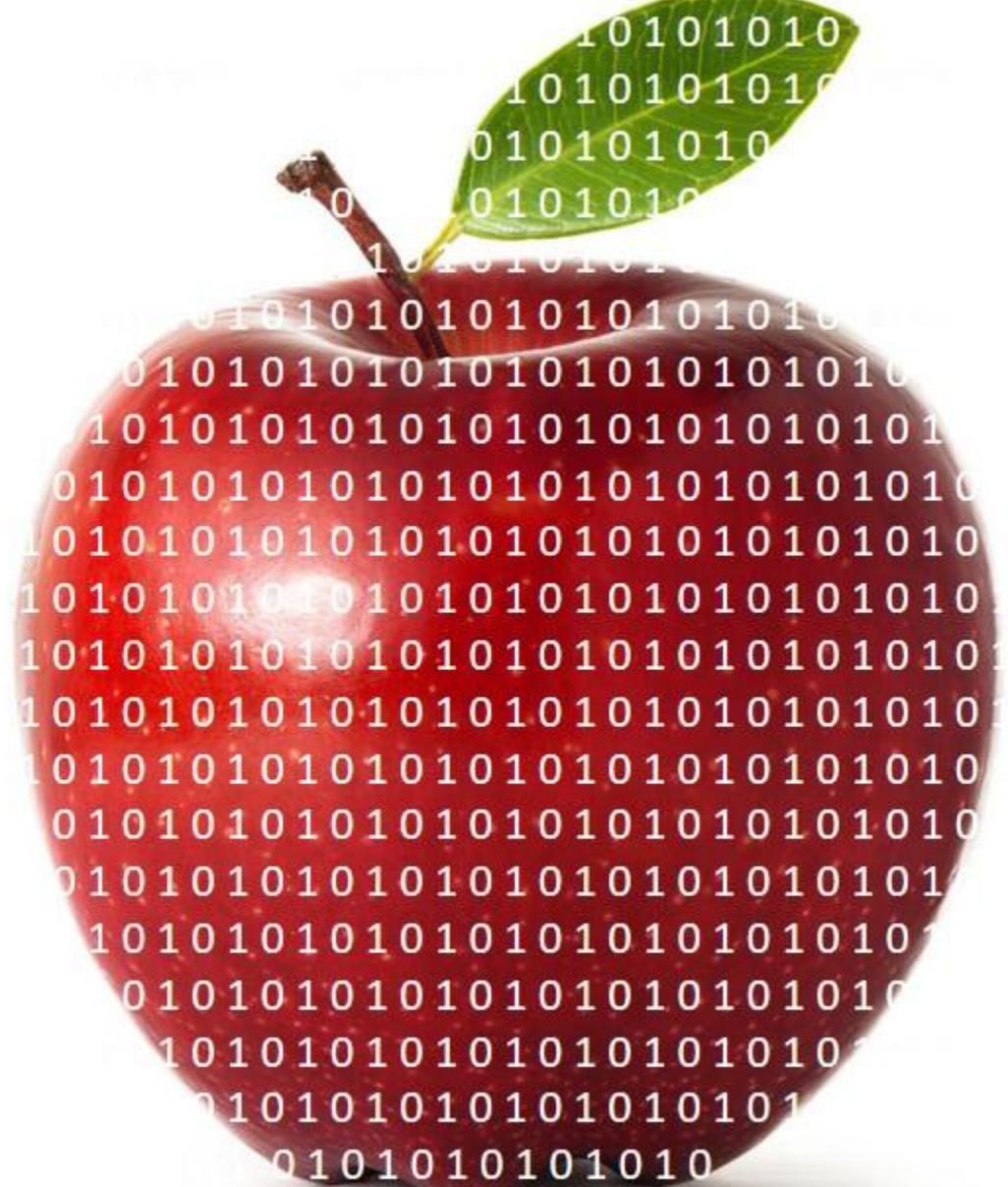
'A shocking exposé ... could make a difference to the way we eat. For ever.'

Evening Standard



What the All-American Meal
is Doing to the World





A red apple with a green leaf, overlaid with binary code.

10101010
1010101010
010101010
10 0101010
101010101
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Five major aspects of Computational Gastronomy Data

1. **Recipes** – Information of recipes that reflects its constituent ingredients.

Further, this could also include details of the temporal sequence of ingredients, the quantity of each ingredient, their state (chopped, ground, etc.), size and the method of processing ingredients (cook, boil, fry, saute, etc.).

Five major aspects of Computational Gastronomy Data

2. Flavor Profiles of Ingredients/Dishes: Every natural ingredient that is used in recipes is primarily selected based on its flavors—taste and odor.

The flavor profile of the ingredient represents the set of empirically reported flavor molecules.

Five major aspects of Computational Gastronomy Data

3. Health Associations of Ingredients/Dishes:

Each ingredient may be associated with the health impacts as identified from experimental studies linking them to diseases.

Five major aspects of Computational Gastronomy Data

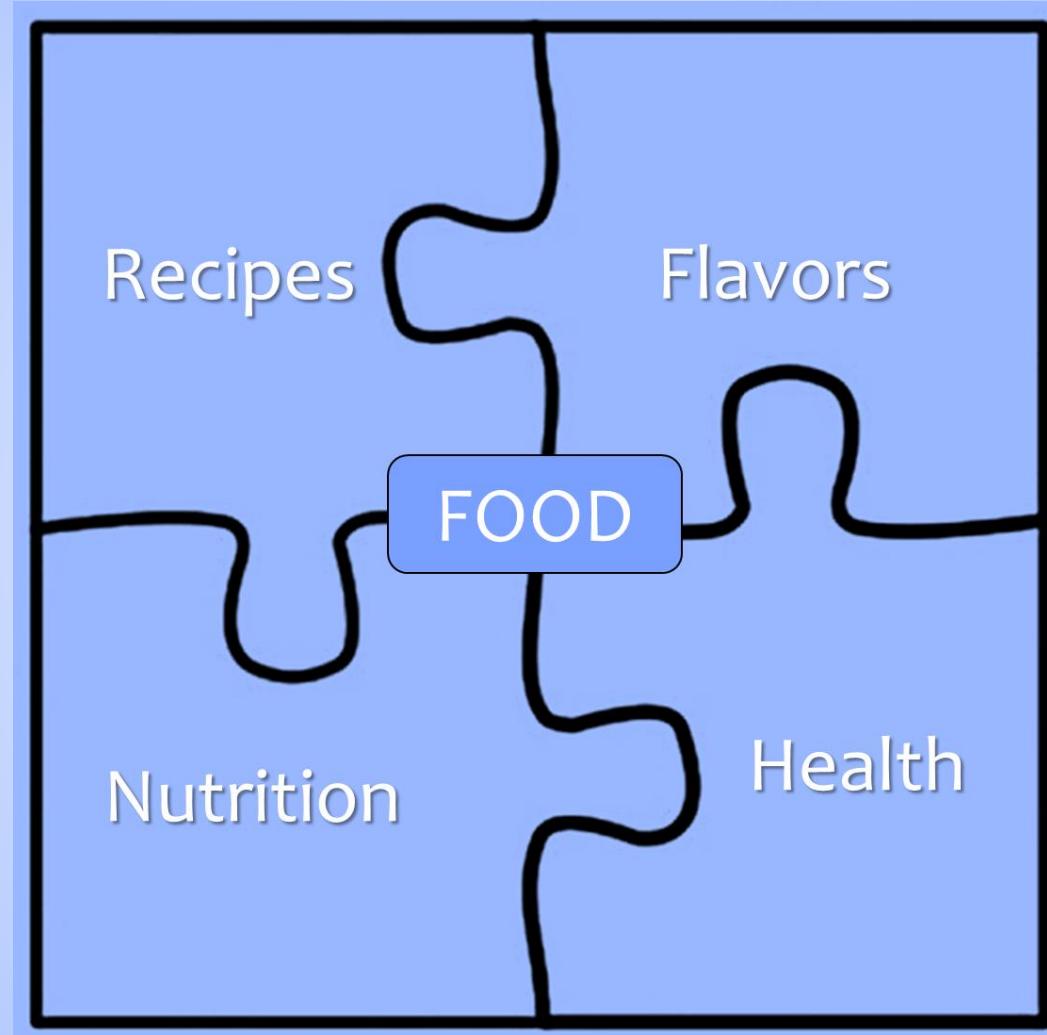
4. Nutritional Profile of Ingredients/Dishes:

Each ingredient can be associated with its nutritional profile, including macro- as well as micro-nutrients.

Five major aspects of Computational Gastronomy Data

5. Carbon Footprints of Food:

Every time we consume food, we leave an ecological footprint on the globe. The carbon footprint of our dietary consumptions consists of around 1/3rd of the total footprint.



Potential Data-Driven Food Innovations

Novel Recipes Design: Cooking is a combinatorial system that combines the raw materials (ingredients) and ways of processing them (cooking rules).

Given that the average recipe uses 10 ingredients and that there are around 1000 unique ingredients, one could create around 10^{30} recipes simply with the power of combinations.

Not every such ‘recipe’ would be palatable, let alone delicious. Knowing the patterns in the traditional recipes from various cuisines, one could filter recipes that conform to cultural norms.

This requires the use of statistics, data/patterns mining, and natural language processing.

Potential Data-Driven Food Innovations

Culinary Fingerprints: Starting with the tripartite data of recipes, their constituent ingredients and further the flavor profiles of ingredients, one could mine for patterns in traditional recipes to obtain the ‘culinary fingerprints’ of cuisines.

These fingerprints represent the characteristic pattern of ingredient co-use as well as the idiosyncratic flavor patterns of a cuisine depicting the culinary preferences of a population.

Such insights are of value for food/recipe design for targeted audiences.

Potential Data-Driven Food Innovations

Taste/Odor Prediction: Large compilation of flavor molecules along with their taste/odor attributes presents an interesting case for machine learning-based taste and odor prediction algorithms.

Here the battery of molecular properties (starting from simple ones such as the number of atoms and molecular weight to all the way up to molecular fingerprints comprising thousands of attributes) are features which could be used for building statistical models for classification into taste and odor classes.

Potential Data-Driven Food Innovations

De novo design of flavor/taste molecules: Along with *in silico* methods for generation of organic compounds, taste/odor classification models can enable de novo synthesis of compounds of desirable profile.

Potential Data-Driven Food Innovations

Healthy and Sustainable Diet Design:

Systematically curated data on the health impacts of ingredients, along with rules for making recipes, can be used for designing potentially healthy recipes.

Interestingly, such strategies could be used to propel the use of ingredients for promoting a sustainable food culture.

Potential Data-Driven Food Innovations

Healthy and Sustainable Diet Design:

Systematically curated data on the health impacts of ingredients, along with rules for making recipes, can be used for designing potentially healthy recipes.

Interestingly, such strategies could be used to propel the use of ingredients for promoting a sustainable food culture.



- Additive
- Animal product
- Beverage
- Cereal/ Crop
- Dairy
- Fish/ Seafood
- Flower
- Fruit
- Herb
- Meat
- Nut/ Seed
- Plant
- Plant derivative
- Pulse
- Spice
- Vegetable

Food: Forms and shapes

Ingredient Categories: Forms and Shapes

- **Vegetable:** All kind of leafy vegetables (spinach, fenugreek), beans (kidney bean, red beans), tubers (potato, turnip, beetroot) and such.
- **Spice:** Flavorful ingredients primarily from root, bark and seeds. Examples: cumin, clove, cardamom, cinnamon, turmeric etc.
- **Herb:** Flavorful ingredients primarily from leaves and stems of plants. Examples: coriander, thyme, rosemary, mint, basil etc.
- **Pulse:** The fruit or seed. Examples: chickpeas, green gram, back gram, pigeon pea etc.

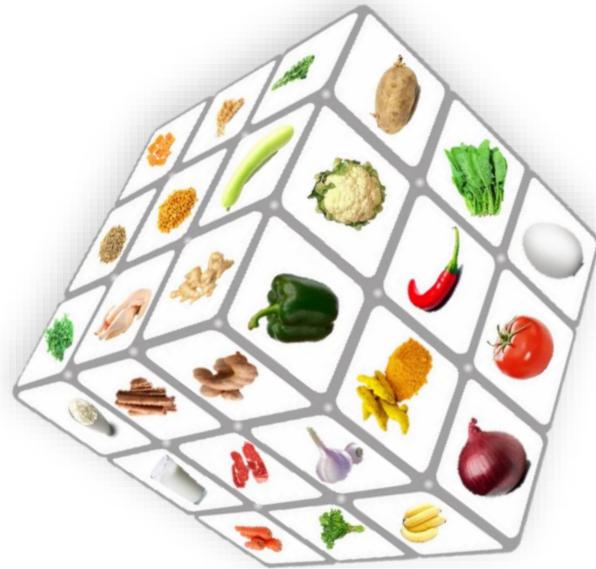
Ingredient Categories: Forms and Shapes

- **Cereal/Crop:** Examples: rice, oat, sorghum, barley etc.
- **Plant:** Examples: tea, thistle, pine, hop
- **Fruit:** Examples: apple, orange, guava, grape, papaya etc.
- **Flower:** rose, jasmine, lavender, artichoke etc.
- **Nut/Seed:** Examples: peanut, walnut, hazelnut, filbert etc.
- **Dairy:** Example: milk, curd/yogurt, butter, buttermilk, cream, cheese, whey etc.

Ingredient Categories: Forms and Shapes

- **Meat:** Examples: beef, pork, mutton, chicken etc.
- **Fish:** Examples: salmon, codfish, tuna, catfish, bluefish etc.
- **Seafood:** Examples: lobster, mollusk, oyster, prawn, shrimp, kombu, kelp etc.
- **Beverage-Alcoholic:** Examples: rum, gin, wine, whisky etc.
- **Beverage:** fruit juice, soft drink, milk shake etc.
- **Additives:** salt, sugar, Monosodium glutamate (MSG)

Computational Gastronomy: Technology & Food



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