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**Plant based eating and health  
outcome: findings from the EPIC  
Study – European perspective**

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**Why to study diet-disease relationships?**



**Aetiology of diseases**

Clear biological meaning  
of the diet factor

Role within the biological  
mechnisms of disease  
formation

Congruent relationships  
with intermediate disease  
markers

**Public Health implications**

Grade of evidence of causal  
relationships

Strength of the diet-disease  
relationship

Attributable risk (prevalence of  
dietary factor)

Chance of change

**Public Health importance of well established diet-disease relationships?**



Establishment of grade of evidence of a causal diet-disease relationship

Evidence grade I*	
Ia	Meta-Analyses of randomized and controlled intervention studies
Ib	Randomized and controlled intervention study
Ic	Non-randomized/ non-controlled intervention study
Evidence grade II	
IIa	Meta-Analyses of cohort studies
IIb	Cohort study

\*according to DGE-Guideline methodology

**Public Health importance of well established diet-disease relationships – conversion into practise**



Food based dietary guidelines - FBDGs)


Scientific process of developing FBDGs (European Food Safety Authority 2010)

Step	
1	<b>Identification of diet-health relationships</b>
2	Identification of country specific diet-related health problems
3	Identification of nutrients of public health importance
4	Identification of foods relevant for food-based dietary guidelines
5	Identification of food consumption patterns
6	Testing and optimising food-based dietary guidelines
7	Graphical representations of food-based dietary guidelines


**Grade of evidence of food-disease relationships on the basis of meta-analysis 2016 (13. Nutrition Report)**

↓↓ probable inverse relation      ↑↑ probable direct relation

Food group	Cancer	Diabetes mellitus Type 2	Coronary heart disease	Stroke
Whole grains	↓↓	↓↓↓	↓↓	↓↓
Vegetables	↓↓	↓	↓↓	↓↓
Fruit	↓↓	↓	↓↓	↓↓
Milk and milk products	↑↓	↓↓	↓	↓
Fermented milk products	◦	↓↓	◦	↓
Red meat	↑↑	↑	◦	↑↑
Processed meat	↑↑	↑↑	↑↑	↑↑
Fish	◦	◦	↓↓	↓↓
Nuts	↓	↓	↓↓	↓↓
Chocolate/cacao	-	-	↓↓	↓↓



**Contribution of prospective observational studies to the diet-disease relationships**



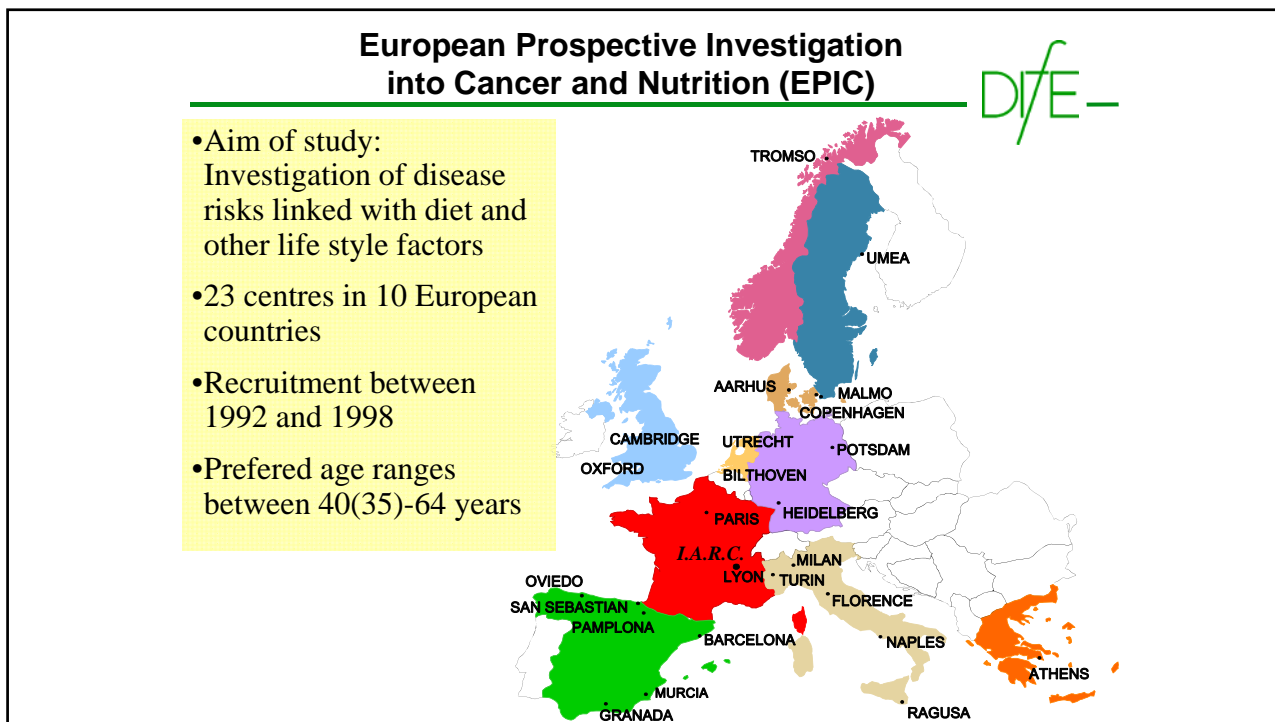
Prospective studies always have a time element (disease endpoints occur only with study time)

Only „old“ studies can make a significant contribution

There are numerous endpoints occurring in a study population

Only some of the endpoints are of prime interest for public health such as

- Mortality and cause specific mortality
- Diseases contributing importantly to mortality such as CVD and cancer
- Metabolic diseases increasing risk for diseases contributing to mortality such as type 2 diabetes
- Metabolic disturbances contributing to metabolic diseases and diseases contributing to mortality such as obesity and hypertension



### EPIC – study participants

Country	Men	Women	Total
Greece	11,954	16,618	28,572
Spain	15,632	25,808	41,440
Italy	15,171	27,516	42,687
France	0	72,996	72,996
Netherlands	10,280	29,792	40,072
Germany	22,833	30,261	53,094
UK	26,912	61,028	87,940
Danemark	27,179	29,875	57,054
Sweden	23,496	30,334	53,830
Norway	0	37,231	37,231
<b>Total</b>	<b>53,457</b>	<b>366,521</b>	<b>519,978</b>

## EPIC data related to lifestyle and personal history



### From questionnaires and interviews:

- Tobacco
- Alcohol
- Reproductive history
- Occupation
- Illnesses
- Physical activity
- Socio-economic status

### From physical examination:

#### Anthropometry

- Height
- Weight
- Waist circumference
- Hip circumference
- Sitting height



#### Blood pressure

#### Blood concentrations (from aliquoted blood stored in liquid nitrogen)

## EPIC data on diet



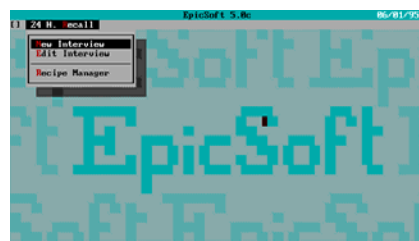
### Two dietary measurements:

Dietary questionnaires on usual diet from all 519,978 subjects

- Very detailed, 150 to 300 foods per questionnaire
- To relate diet to disease risk

One day "actual" diet from a 7% sample of subjects (38,000)

- Computerized, 3000 foods and 700 recipes per country
- To calibrate dietary measurements between countries





**EPIC results regarding food groups associated with disease risk**

**D/E-**

**Grade of evidence of at least probable (according to nutrition report 2016):**

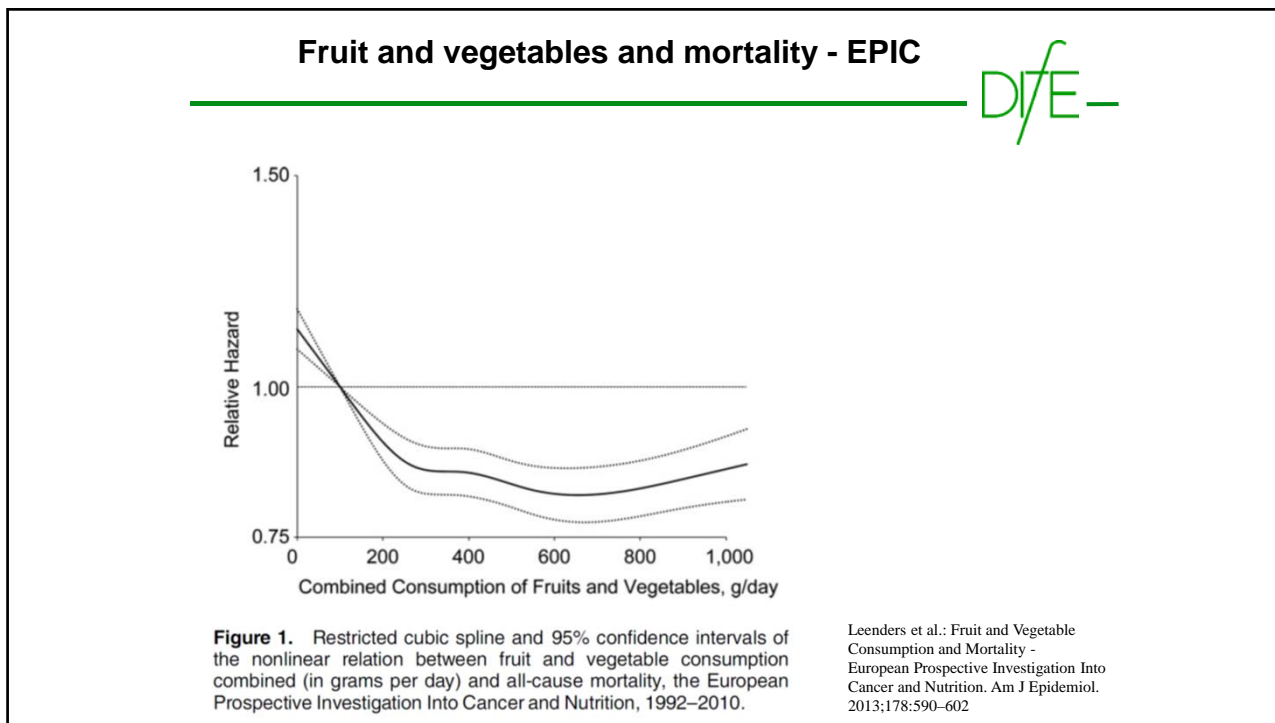
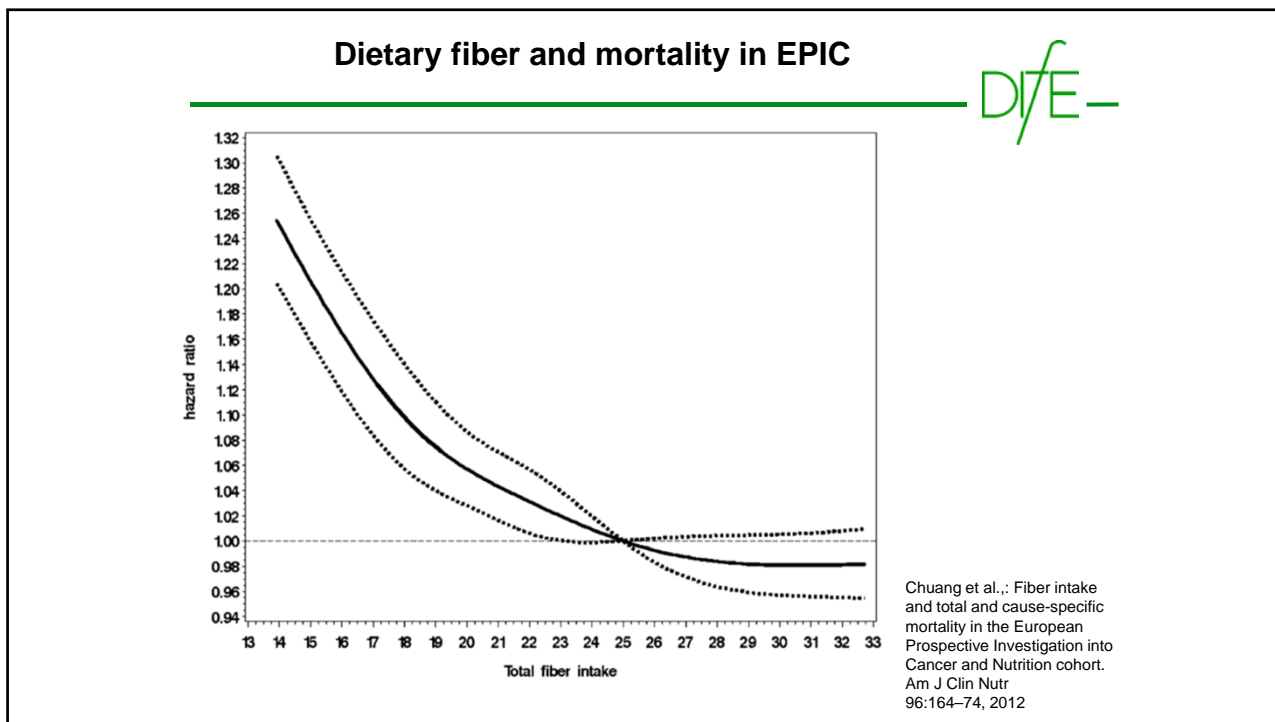
**Plant based**

- Whole grain foods (dietary fiber)
- Vegetable
- Fruits
- Nuts
- Chocolate/Cocoa

**It is important to see the results for different endpoints based on identical dietary data (and the same blood aliquotes)**

**Animal based**

- Milk and milk products/fermented milk products
- Fish
- Red meat/Processed meat



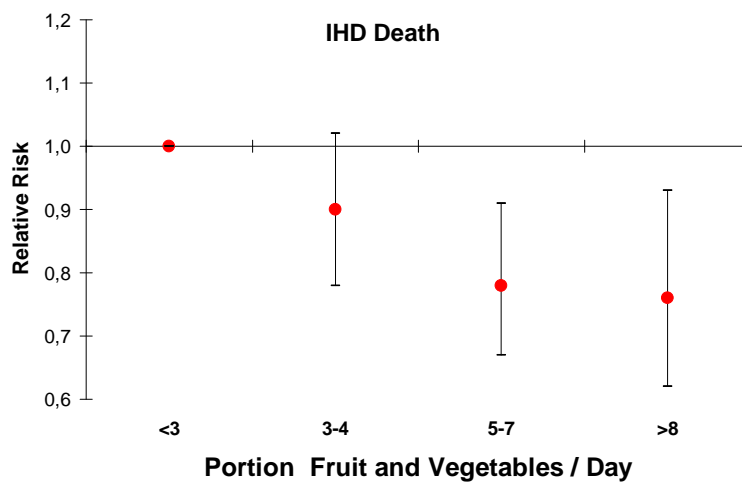
**Fruit and vegetable quintiles and risk of cancer - EPIC**



	Quintiles of intake of fruit and vegetables (g/d)				
	Q1	Q2	Q3	Q4	Q5
<b>Relative Risk</b>	<b>1</b>	<b>0.95</b> (0.92-0.99)	<b>0.91</b> (0.88-0.95)	<b>0.93</b> (0.89-0.97)	<b>0.89</b> (0.85-0.93)

Boffetta et al., JNCI 102, 529-537: 2010

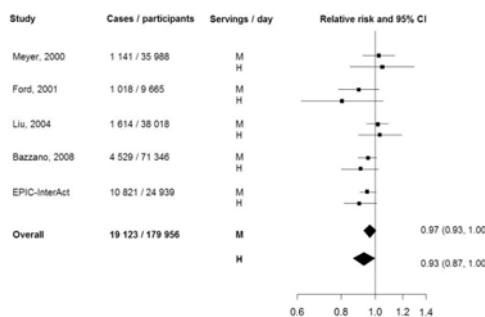
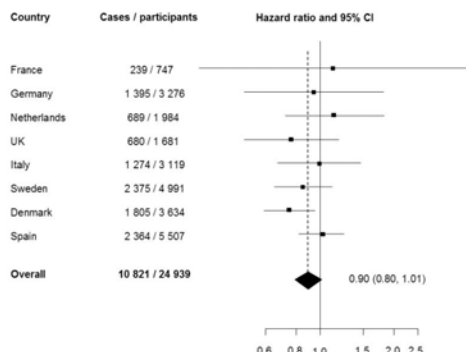
**EPIC-Heart study on mortality**



Crowe et al., European Heart Journal 2011, 32:1235-1243



### Fruit and vegetables and type 2 diabetes - EPIC



Cooper et al.: Fruit and vegetable intake and type 2 diabetes: EPIC-InterAct prospective study and meta-analysis. Eur J Clin Nutr. 2012; 66: 1082-1092.

### Polyphenols as marker of food intake

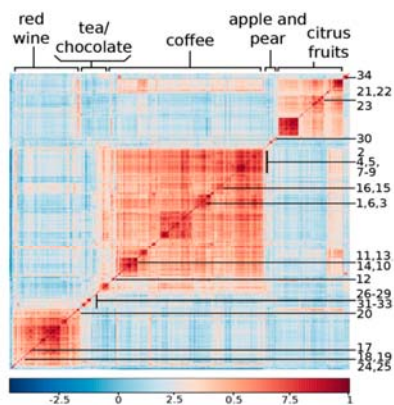


FIGURE 3 Metabolite-metabolite correlation analysis of the 2272 mass spectral features correlated to dietary intakes of 6 polyphenol-rich foods (24-h dietary recalls). The heat map shows clusters for each food made of mass spectral features derived from metabolites co-occurring in a same food and excreted in the same urine samples. Numbers refer to the annotated polyphenol metabolites (Table 1).

William MB Edmands, Pietro Ferrari, Joseph A Rothwell, Sabina Rinaldi, Nadia Slimani, Dinesh K Barupal, Carine Biessy, Mazda Jenab, Françoise Clavel-Chapelon, Guy Fagherazzi, Marie-Christine Boutron-Ruault, Verena A Katzke, Tilman Kühn, Heiner Boeing, Antonia Trichopoulou, Pagona Lagiou, Dimitrios Trichopoulos, Domenico Palli, Sara Grioni, Rosario Tumino, Paolo Vineis, Amalia Mattiello, Isabelle Romieu, and Augustin Scalbert: Polyphenol metabolome in human urine and its association with intake of polyphenol-rich foods across European countries. Am J Clin Nutr 2015;102:905-13

## Polyphenols as marker of food intake



Raul Zamora-Ros, David Achaintre, Joseph A. Rothwell, Sabina Rinaldi, Nada Assi, Pietro Ferrari, Michael Leitzmann, Marie-Christine Boutron-Ruault, Guy Fagherazzi, Aurélie Auffret, Tilman Kühn, Verena Katzke, Heiner Boeing, Antonia Trichopoulou, Androniki Naska, Effie Vasilopoulou, Domenico Palli, Sara Grioni, Amalia Mattiello, Rosario Tumino, Fulvio Ricceri, Nadia Slimani, Isabelle Romieu & Augustin Scalbert: Urinary excretions of 34 dietary polyphenols and their associations with lifestyle factors in the EPIC cohort study. Scientific Reports | 6:26905, 2016

Food	Consumers (n)	Polyphenol (Spearman correlation coefficient)
Red wine	121	Galic acid ethyl ester (0.69), resveratrol (0.59), gallic acid (0.48), hydroxytyrosol (0.43), tyrosol (0.36), (+)-catechin (0.34), <i>p</i> -coumaric acid (0.27), 4-hydroxyphenylacetic acid (0.19), 3,4-dihydroxyphenylacetic acid (0.15)
Coffee	410	Caffeic acid (0.65), protocatechuic acid (0.60), ferulic acid (0.58), <i>m</i> -coumaric acid (0.53), 3,4-dihydroxyphenylpropionic acid (0.51), 3-hydroxybenzoic acid (0.39), vanillic acid (0.31)
Tea	117	Galic acid (0.38), (-)-epicatechin (0.30), (+)-catechin (0.22), quercetin (0.19)
Chocolate	111	(-)-Epicatechin (0.22), vanillic acid (0.15)
Citrus fruits	185	Hesperetin (0.60), naringenin (0.56), kaempferol (0.33)
Citrus juices	151	Hesperetin (0.15), naringenin (0.15), kaempferol (0.10)
Apple and pear	226	Phloretin (0.40), (-)-epicatechin (0.20), 3,4-dihydroxyphenylacetic acid (0.19), homovanillic acid (0.16)
Berries	42	<i>p</i> -Coumaric acid (0.20), (+)-catechin (0.19)
Onion, garlic	220	Quercetin (0.17), apigenin (0.11), isorhamnetin (0.10)
Olive oil	238	Hydroxytyrosol (0.36), tyrosol (0.31), 3,4-dihydroxyphenylacetic acid (0.17), apigenin (0.17)
Olives	44	Hydroxytyrosol (0.34), 3,4-dihydroxyphenylacetic acid (0.29), homovanillic acid (0.22), tyrosol (0.11)
Bread, non-white	260	3,5-Dihydroxybenzoic acid (0.45), 3,5-dihydroxyphenylpropionic acid (0.43), enterolactone (0.25), daidzein (0.20), enterodiol (0.20), genistein (0.19), <i>m</i> -coumaric acid (0.16), ferulic acid (0.13)
Breakfast cereals	32	3,5-Dihydroxybenzoic acid (0.17), 3,5-dihydroxyphenylpropionic acid (0.16), daidzein (0.15), equol (0.08), enterolactone (0.08)
Soya products	9	Genistein (0.17), daidzein (0.10)

**Table 3. Urinary polyphenols most highly correlated to recent food intake in the EPIC cohort.** The top two to nine polyphenols (out of 34 measured polyphenols) most highly correlated with the intake of each food group are listed. The number of reported correlations for each food group was based on current knowledge on polyphenol food composition and polyphenol metabolism. Some additional polyphenols may also be correlated to intake of each food, but they were excluded if not known as a component of the food considered or as a possible metabolite derived from a component of this food.

## Summary and conclusion



- The interest in food-disease relationships is high due to their potential important public health implications
- A prospective cohort is accruing person time of follow up regarding disease endpoints with advancement of time and is contributing study results to grade the evidence of a causal relation between a dietary factor and a disease endpoint
- EPIC is one of the few prospective cohort studies that could look back to a follow up of more than 20 years and a wide range of endpoints.
- The EPIC published about food groups of high public health interest and will have the perspective to exploit in addition to the interview data also the blood samples