

# **Unleashing the power of data through organization**

## **Structure and connections for meaning, learning, and discovery**

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See the full paper for detail and references

# The Future of Knowledge Organization

*Knowledge organization is needed everywhere*

**Create** the future of KO

**Think BIG.** Think answers not pointers.  
Focus on substantive data

**Many areas, tasks, and functions**  
that could profit from KO principles

Engage with Ontologies, AI, data modeling

# Areas, tasks, and functions

- 1 Knowledge bases for question-answering and cognitive systems
- 2 Knowledge base for information extraction from text or multimedia
- 3 Linked data
- 4 Big data and data analytics. Data interoperability and reuse
- 5 Interoperability of operational information systems.  
Electronic health records (EHR) as an example
- 6 Information systems in the enterprise
- 7 Influence diagrams (causal maps), dynamic system models,  
process diagrams, concept maps, and other node-link diagrams
- 8 Knowledge organization for understanding and learning
- 9 Knowledge transfer between domains

# Unification

- **across applications**
- **across types of data** (example: organization database treated like classification)
- **across disciplines**, supports knowledge transfer from one discipline domain to another
- **across languages** (precise definitions)
- **across cultures**, across organizations (organizational cultures)
- **across worldviews**

# **Part 2**

## **The application of Knowledge Organization**

# **2.1 Knowledge bases for question-answering and cognitive computing**

Knowledge base	Some KOS used
<b>CYC</b> Common sense knowledge	CYC Ontology, including entity types, relationship types, and entity values
<b>IBM Watson</b> Custom KB for applications	An extensible inventory of relationship types
<b>Google Knowledge Graph</b> Huge database of varied kind of data (Starr 2014)	schema.org for entity types and relationship types
<b>DBpedia</b> Large database of statements extracted from Wikipedia	DBpedia Ontology (E-R schema) Authority lists for individual entity values (instances), each identified by a URI.
<b>GDELT</b> Event reports	CAMEO Coding Scheme for events Own list of 300 themes, World Bank Taxonomy themes 2,300 emotions and themes (from 24 sentiment analysis packages) US government geonames standards

## **2.2 Knowledge base for information extraction from text or multimedia**

Often only text is considered, but information can be extracted from graphs and video (for example, identifying people by face recognition and relationships between people from analyzing scenes). In the following text+



# Information extraction

- **Entity extraction** (Named-entity recognition)  
Locating references to entities in text+, associate with a unique identifier.
- **Information extraction**  
Formally represent the propositions the text makes about these entities.

Information extraction both uses and feeds knowledge bases for question answering.

# KOS for information extraction

Information extraction needs much knowledge, which must be properly organized into KOS

- **Linguistic knowledge**: morphological, part-of-speech, and lexical (meaning). Lexicalized phrases.
- **Large KOS listing entity values** and their (multiple) names (persons, organizations, places, concepts/subjects, ...)
- **Knowledge supporting word sense disambiguation** (WSD). Both linguistic knowledge and world knowledge.

## 2.3 Linked data

- **Entity-relationship data model**
- **Data from independent data sets can linked**
- **Key implementation component of the Semantic Web**
- **Enormous opportunity for KO.**
  - Deploying KOS data on the Web and have them more widely used.
  - Linked data require properly structured and often very large KOS.

# Linked data

- **The more pervasive **standardization**** with respect to
  - entity types
  - relationship types
  - entity values**the more successful linked data searching will be**
- **This is a **problem of knowledge organization****

**Drug <hasName> Text**

**Drug <hasGenericVersion> Drug**

**Drug <hasActiveIngredient>  
ChemicalSubstance**

**Drug <hasClinicalPharmacologyDescr>  
Text**

**Drug <hasIndicationDescr> Text**

**Drug <hasContraIndicationDescription>  
Text**

**Drug <administeredVia>  
RouteOfAdministration**

**DBDrug <hasName> Text**

**DBDrug <hasGenericName> Text**

**DBDrug <hasCASRegistryNumber> URI**

**DBDrug <hasAbsorptionDescr> Text**

**DBDrug <hasBioTransformDescr> Text**

**DBDrug <hasPharmacolDescr> Text**

**DBDrug <hasProteinBindRate> Pct**

**DBDrug <hasIndicationDescr> Text**

**DBDrug <hasPossibleDiseaseTarget>  
Disease**

**DBDrug <hasContraIndicationInsert>  
Document**

**DBDrug <hasDosageForm> DosageForm**

# **2.4 Big data and data analytics. Data interoperability and reuse**

# Example 1. Merging like datasets

- **Research question:** Factors affecting school success
- **Need large sample, so merge data sets** with anonymized data on individual students and test scores from many US states (many European countries)
- **Problem:** this **works only if variables are defined the same way in all data sets**
  - Factors such as *socio-economic status of the student or home environment*
  - Concepts and skills covered in the tests.
- **This is a knowledge organization problem**

# Example 2. Linking datasets

- **Research question:** relationships between
  - *per capita income,*
  - *how people feel about the economy,* and
  - *birth rate*

Unit of analysis: Locality

- **The variables needed are in three different data sets:**
  - 1 per-capita income by locality
  - 2 Twitter messages (analyze for sentiment)
  - 3 Birth rate by locality

**The data sets need to be linked** so that for each locality we have values for the three variables

- **Problem:** The **ability to link these data sets depends on the linking variable, locality, being defined the same way** and identifiable (a problem with Twitter)



## 2.5 Interoperability of operational information systems

### Electronic health records (EHR) as an example

- Interoperability of EHR data is an **obvious must**, but far from solved.
- **Needs KOS for**
  - race/ethnicity, age, sex
  - bodily or mental functions or conditions
  - diseases
  - medical procedures
  - drugs
- **Worked on heavily, mainly by people in biomedical informatics / biomedical ontologies.**
- Given here as one example of the importance of KO for operational systems.

## **2.6 Information systems in the enterprise**

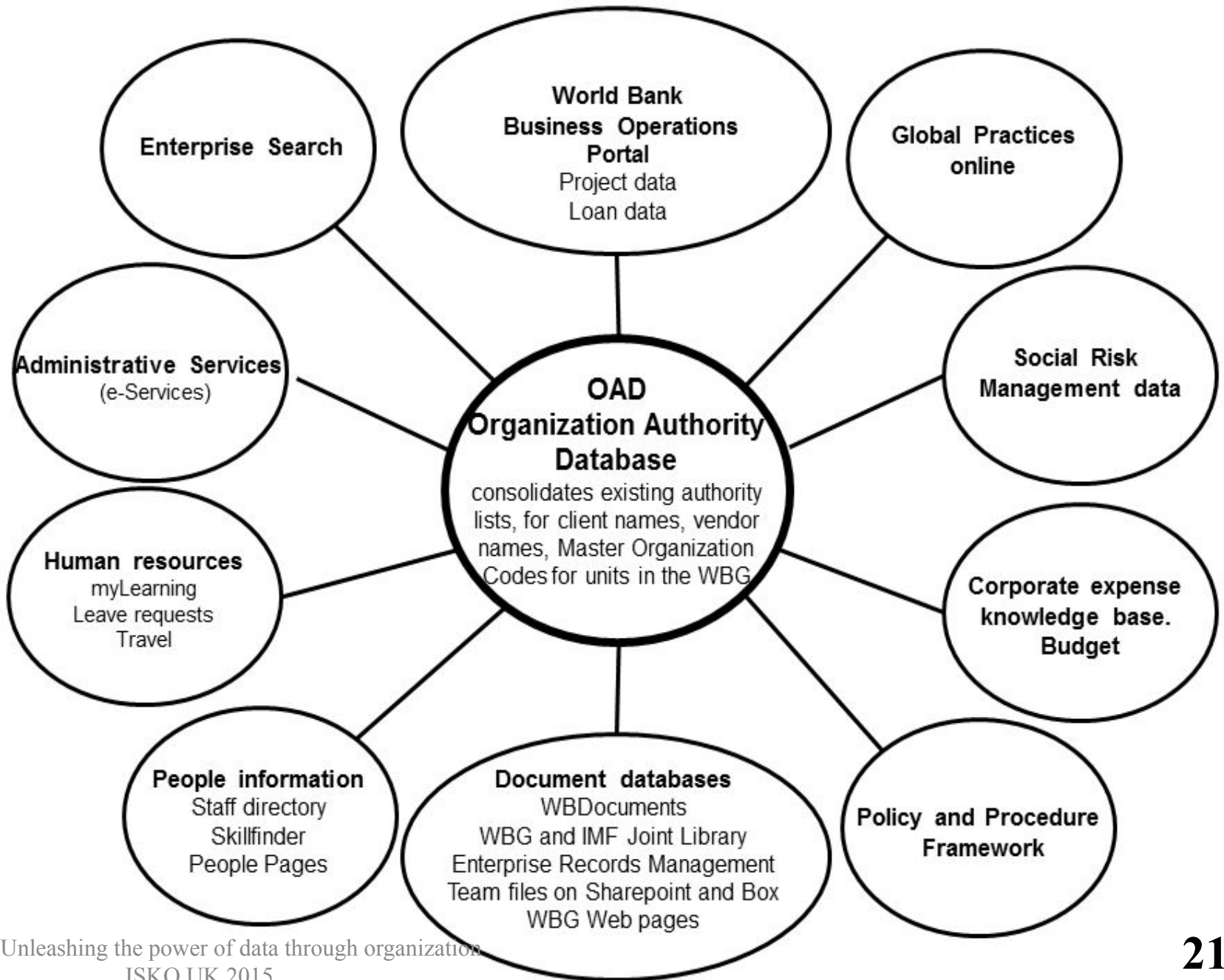
# Example 1

- **Problem:**  
Many organizations do not know in a central place what data they have
- **Solution:**
  - **Develop an enterprise-wide entity-relationship conceptual data schema** (an enterprise ontology, an enterprise data model, the modern version of a data dictionary), using ideas from Web standards.
  - Use this to **organize an inventory or registry of all data systems** in the organization and the specific pieces of data in each.

# Example 2

## Unified authority database for Organizations

considered for the World Bank Group (WBG)



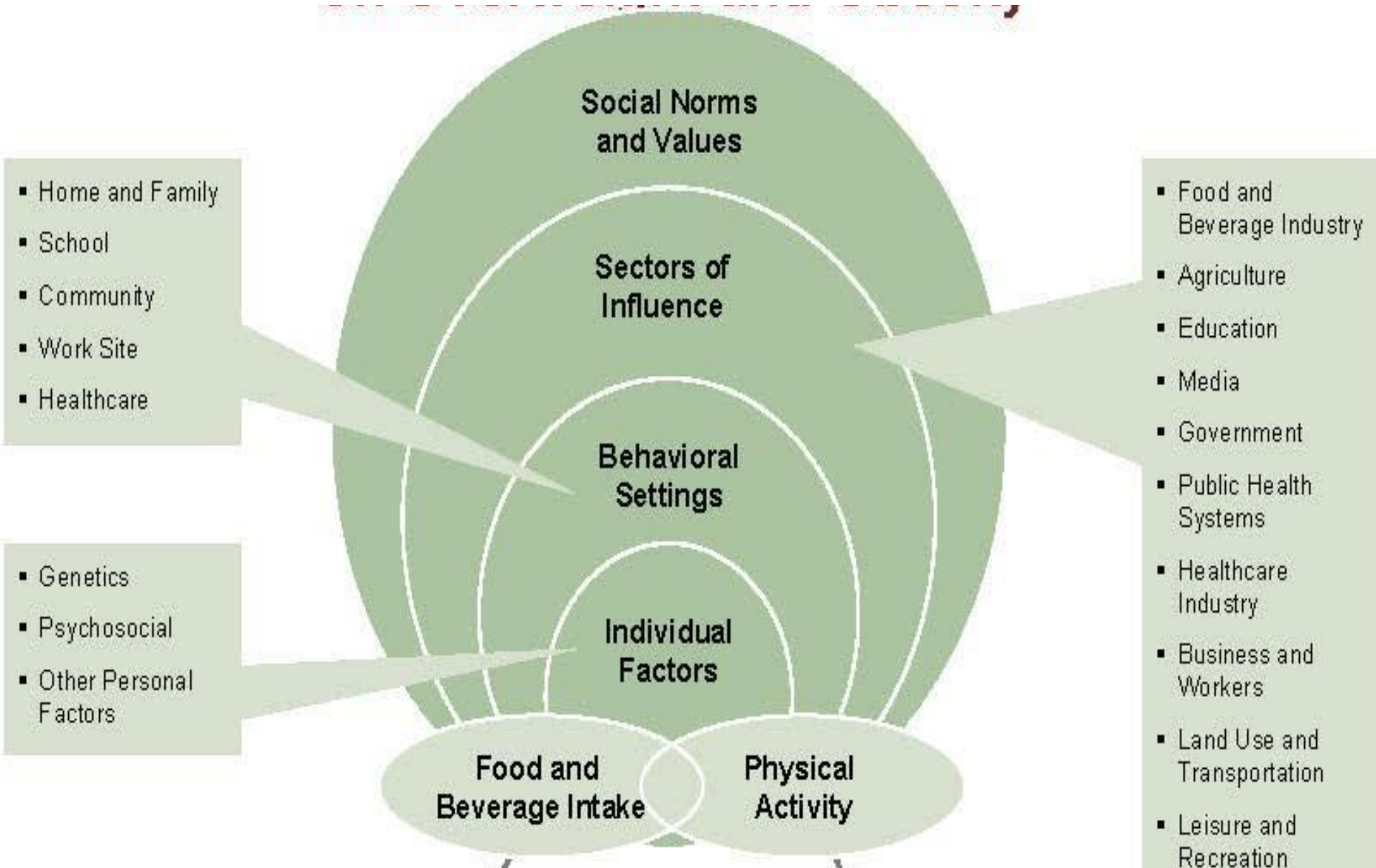
# Example 2 cont.

- The enterprise-wide Organization Authority Database should be structured exactly like a hierarchical thesaurus: Just like concepts, the organizations form a hierarchy, and they have multiple names

# 2.7 Node-link diagrams

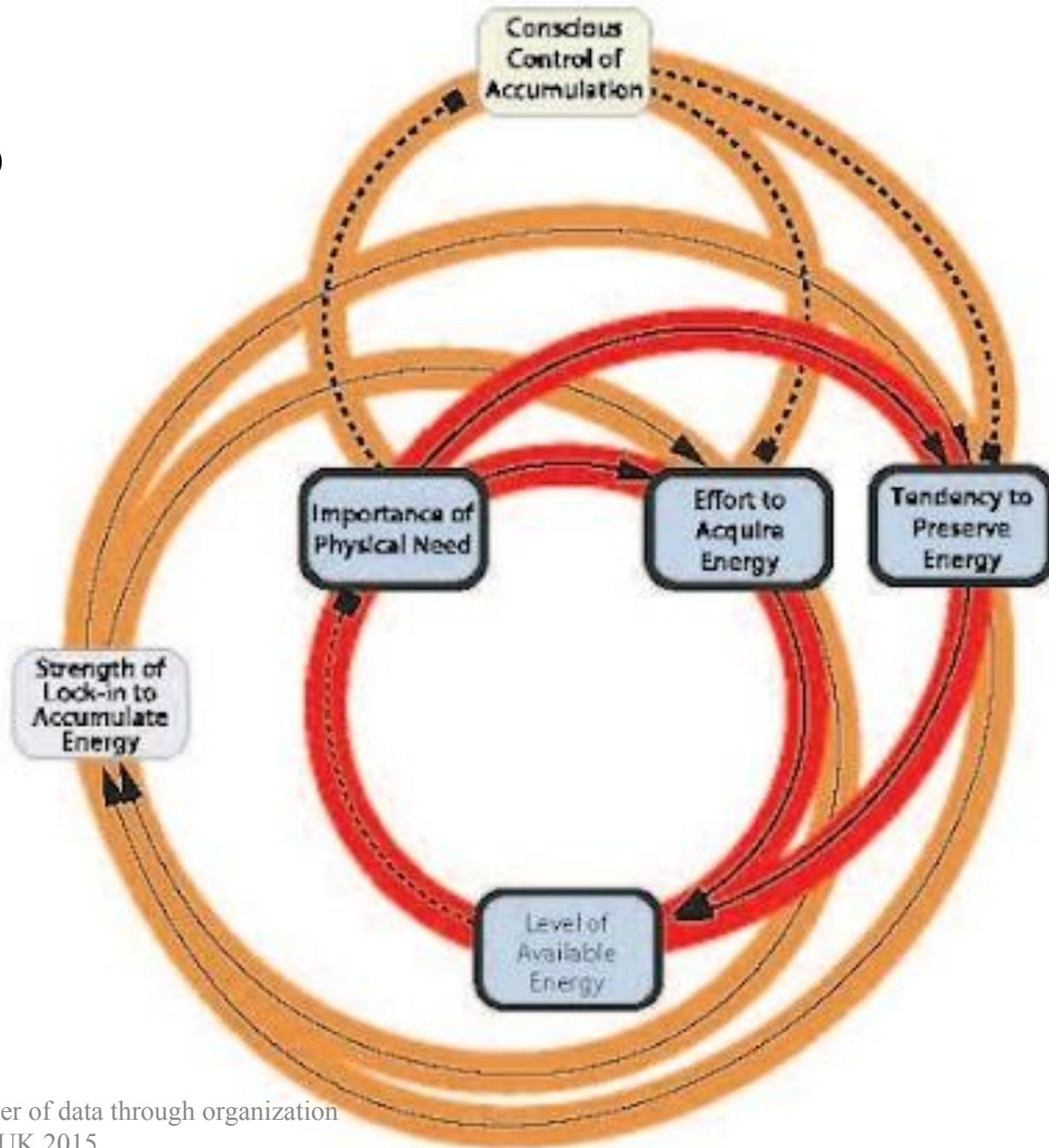
- **Causal maps (influence diagrams)**
- **Dynamic system models**
- **Process diagrams**
- **Concept maps**
- **Other**

# Influences on overweight and obesity

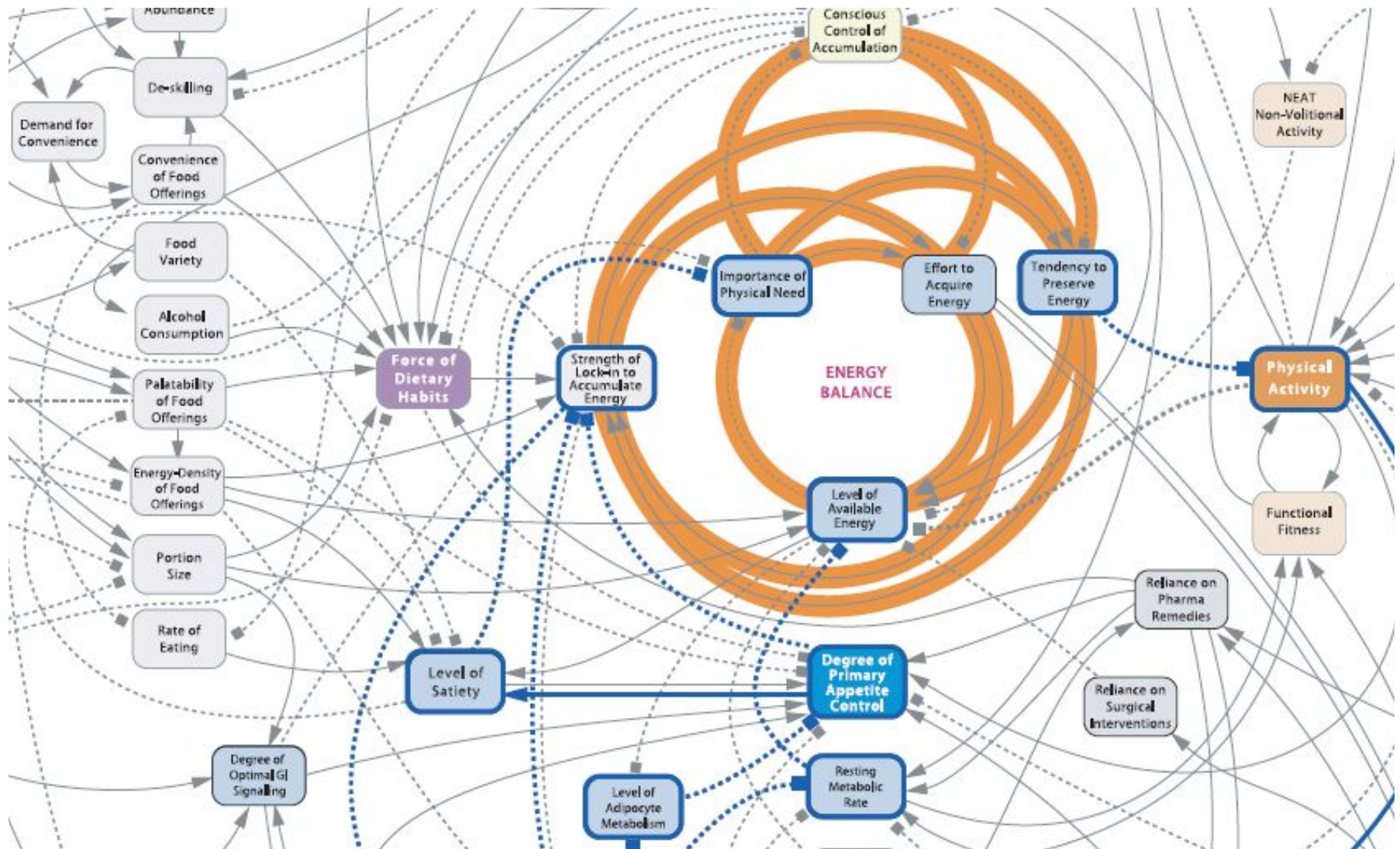




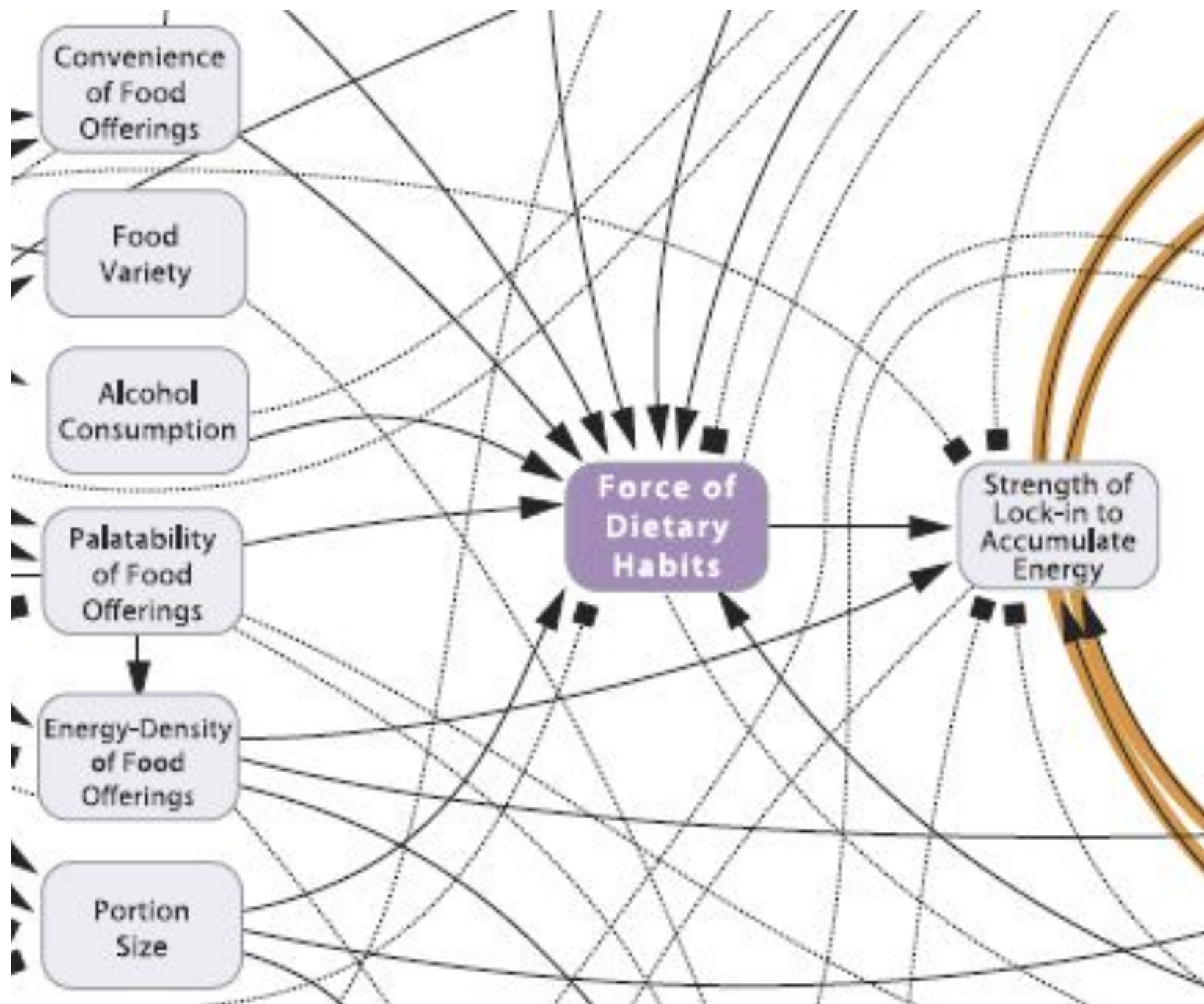
# shiftN causal map for obesity



# Segment the large and detailed shiftN causal map for obesity









# KO issues

- Arranging variables in a meaningful order
- **Mapping variables from one model to another**

Coming up later

- **Merging node-link diagrams**
- **Linking node-link diagrams**

**shiftN causal map variables. Top level with example detail (arranged by DS)**

Individual	Environment
<p><b>Engine</b></p> <ul style="list-style-type: none"> <li>Energy balance</li> <li>Conscious control of accumulation</li> <li>Effort to acquire energy</li> <li>Strength of lock-in to accumulate energy</li> </ul>	
<p><b>Physiology</b></p> <ul style="list-style-type: none"> <li>Degree of primary appetite control by brain</li> <li>Genetic and/or epigenetic predisposition</li> </ul>	
<p><b>Food consumption</b></p> <ul style="list-style-type: none"> <li>Force of dietary habits</li> <li>Tendency to graze</li> <li>Demand for convenience</li> <li>Food exposure</li> <li>Food variety</li> </ul>	<p><b>Food production</b></p> <ul style="list-style-type: none"> <li>Societal pressure to consume</li> <li>Demand for health</li> <li>Pressure to improve access to food offerings</li> <li>Cost of ingredients</li> </ul>
<p><b>Individual physical activity</b></p> <ul style="list-style-type: none"> <li>Level of transport activity</li> </ul>	<p><b>Physical activity environment</b></p> <ul style="list-style-type: none"> <li>Dominance of motorised transport</li> <li>Opportunity for unmotorised transport</li> </ul>
<p><b>Individual psychology</b></p> <ul style="list-style-type: none"> <li>Food literacy</li> <li>Stress</li> </ul>	<p><b>Social psychology</b></p> <ul style="list-style-type: none"> <li>Exposure to food advertising</li> <li>Peer pressure</li> </ul>

# **Some (approximate) matches and non-matches between 4 lists of variables**

<b>shiftN</b>	<b>Kaplan</b>	<b>Nanotechnology</b>	<b>Downey' list</b>
<b>Engine</b>			
<b>Energy balance</b>	<b>Energy balance</b>		
	<b>Energy intake</b>		
	<b>Energy expenditure</b>		
<b>Conscious control of accumulation</b>			<b>lack of self-control</b>
<b>Effort to acquire energy</b>			
			<b>Response to food cues</b>
<b>Physiology</b>			
<b>Appetite control by brain</b>			
<b>Genetic &amp; epigenetic predisposition</b>			<b>genetics epigenetic factors</b>
<b>Food consumption</b>	<b>Food and bev. intake</b>		<b>overeating</b>
<b>Force of dietary habits</b>			
		<b>Malnutrition (conv. foods)</b>	<b>high fruct. corn syrup</b>
<b>Food production</b>	<b>Food &amp; bev. industry</b>	<b>Agricultural production</b>	<b>agricultural policies</b>
		<b>Food deserts</b>	<b>food deserts</b>
<b>Cost of ingredients</b>			
<b>Indiv. physical activity</b>	<b>Physical activity</b>	<b>Exercise &amp; physical activity</b>	
		<b>Lack of exercise</b>	<b>Low physical activity</b>



# More uses of node-link diagrams

## In biology and in industrial engineering

- diagrams of sequential and interrelated processes that lead to some outcome or state

## In biology

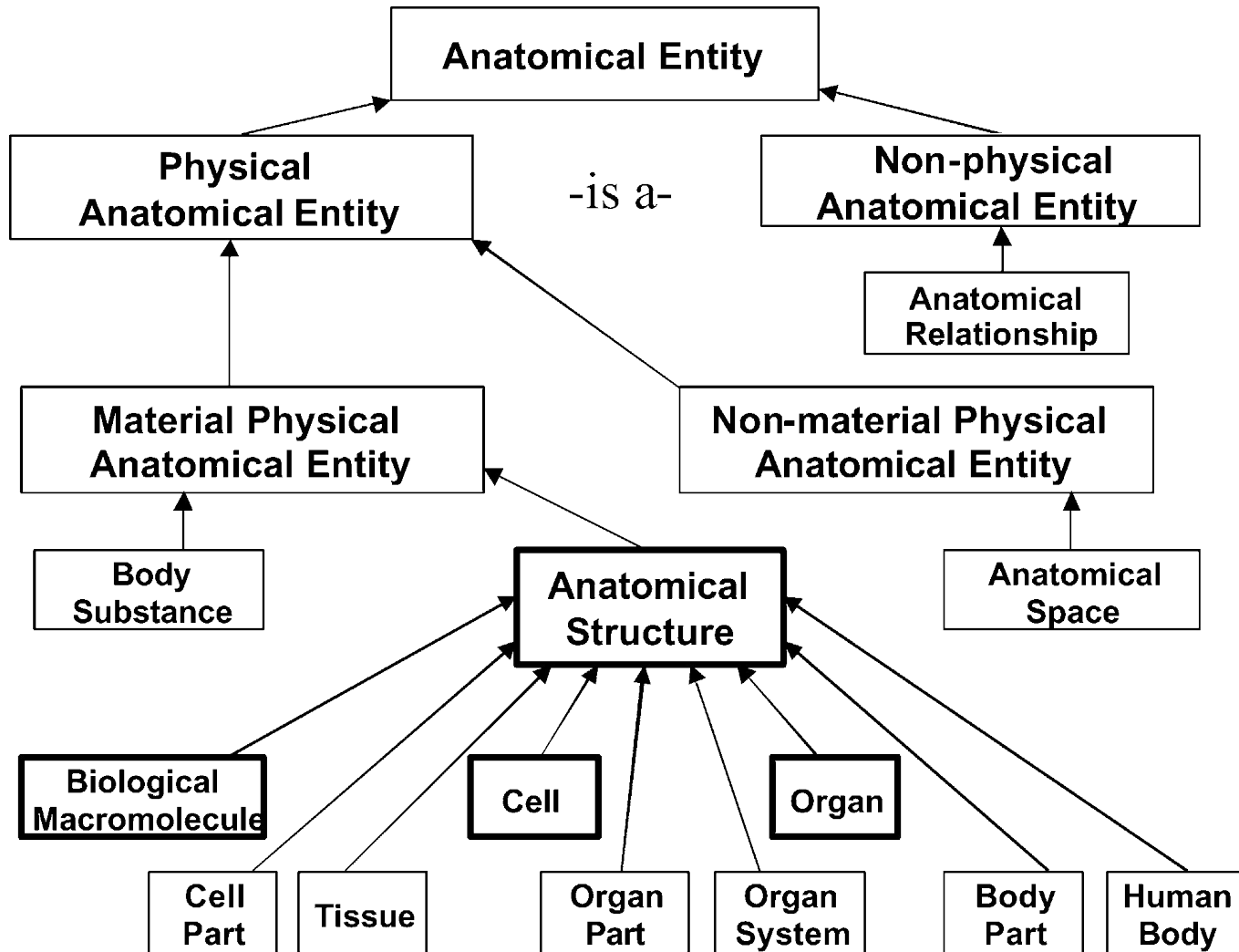
- diagrams of signaling pathways,
- diagrams of metabolic networks,
- diagrams of gene regulatory networks

# Concept maps

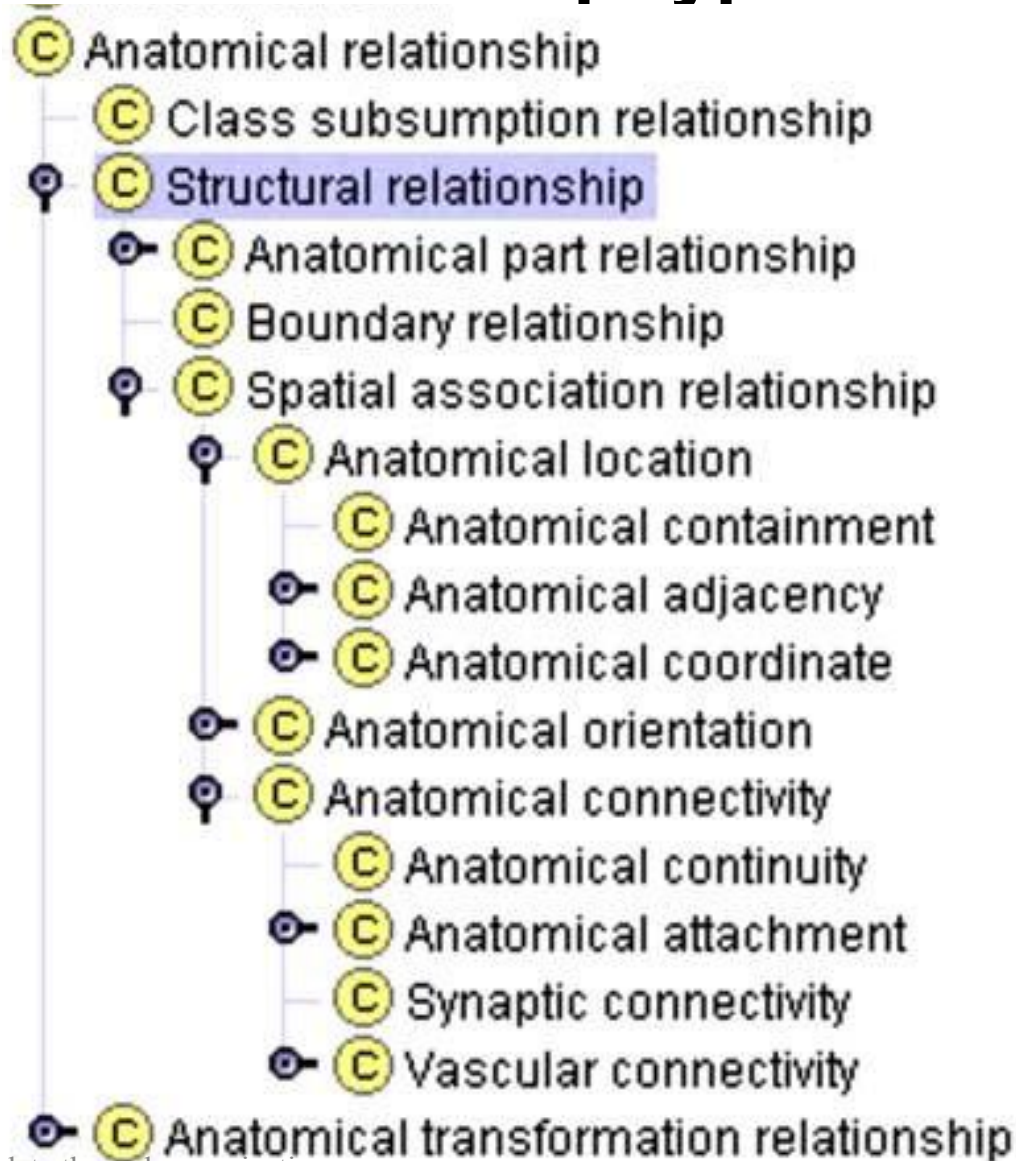
- Used as thesaurus displays since the 1950s
- Resurfaced forcefully in education
- If you know of earlier uses, let me know

# **2.8 Knowledge organization for understanding and learning**

# Foundational Model of Anatomy: Entity types



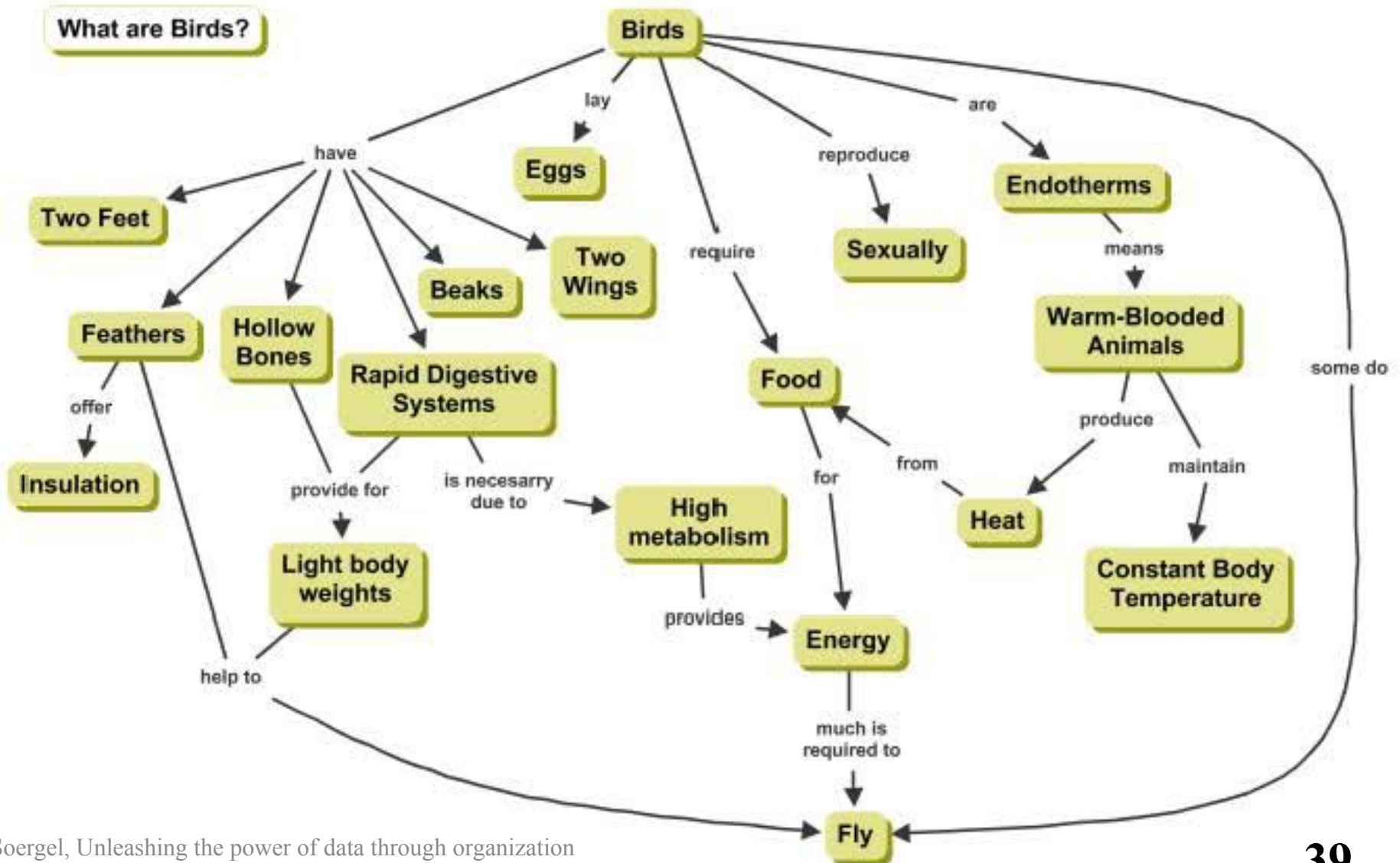
# Foundational Model of Anatomy: Relationship types



## Hypothesis

Students who are taught anatomy using the *Foundational Model of Anatomy* have a better grasp of the structure of the body.

# Concept map about birds



# Concept map hypotheses

The bird concept map will allow learners to form a better internal representation of a bird as a system.

Constructing concept maps will help learners to develop a better understanding (a better structured mental model) of the topic.



# Meaningful arrangement

## Classification of animals

### a. Britannica Elementary: Menu for *Animal Kingdom*

Thoughtless arrangement, devoid of any meaning  
One of many such examples from children's websites

### b. Meaningful arrangement

Based on the modern cladogram shown in c.

### c. Vertebrates cladogram

# Britannica Elementary: Menu for *Animal Kingdom*

Thoughtless arrangement, devoid of any meaning

## Explore by Group



Arthropods



Birds



Mammals



Mollusks



Fish



Reptiles and Amphibians



Prehistoric Life



Other Interesting Animals

# Animal Kingdom: Meaningful arrangement based on modern science

## Animals without a spine (*invertebrates*)

Snails, octopus, mussels (*mollusks*)

Bugs (insects), spiders, crabs (*arthropods*)

## Animals with a spine (*vertebrates*)

### *Fish*

Frogs, toads, salamanders (*amphibians*)

Lizards&snakes, crocodiles, dinosaurs, birds

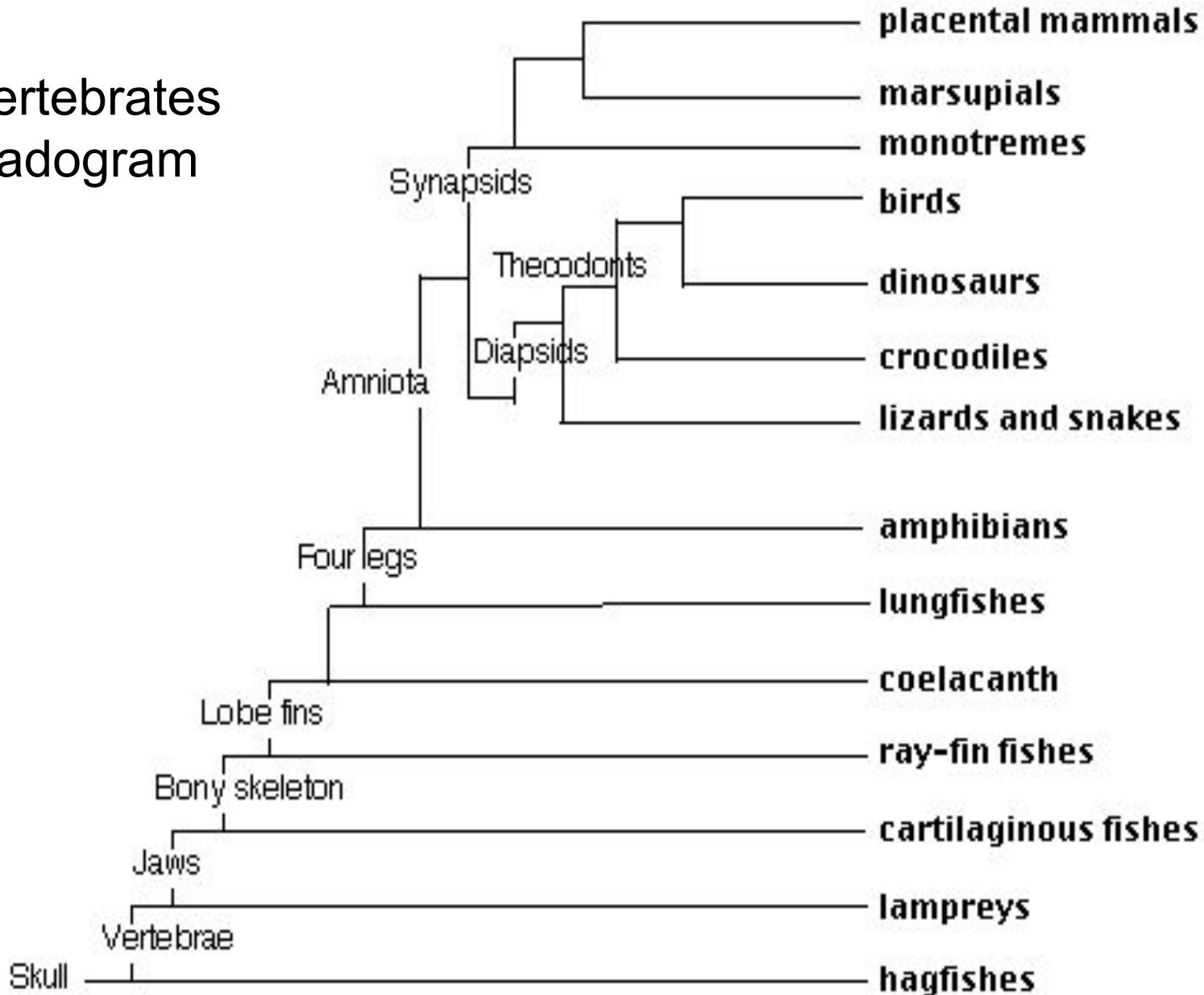
Lizards&snakes, crocodiles, dinosaurs (*reptiles*)

### *Birds*

Elephants, whales, cows, dogs, bats, mice, monkeys, apes,  
humans (*mammals*)

Note: Could simplify, add pictures

# Vertebrates cladogram



# Meaningful arrangement hypothesis

Young students who use the animal home page with the meaningful arrangement will over time absorb the sequence and perceive a progression. When much later in biology the structure of the animal kingdom and the evolution of animals are discussed, these students will understand more quickly.

# 2.9 Knowledge transfer between domains

## Management styles and educational styles compared

Style of social interaction	Management style	Educational style
<b>Autocratic, authoritarian, directive</b>	<b>Autocratic, authoritarian, directive (coercive), top-down</b>	<b>Direct instruction, teacher-centered Teacher as formal authority, expert</b>
<b>Military style</b>	<b>Military style</b>	<b>Military style</b>
<b>Paternalistic</b>	<b>Paternalistic</b>	
<b>Authoritative (visionary)</b>	<b>Authoritative (visionary)</b>	
<b>Persuasive</b>	<b>Persuasive</b>	
<b>Coaching</b>	<b>Coaching</b>	<b>Teacher as facilitator</b>
<b>Individual inner discipline, motivation, agreement with norms</b>		<b>Montessori</b>

**Figure 17. Management styles and educational styles compared**

<b>Style of social interaction</b>	<b>Management style</b>	<b>Educational style</b>
<b>Participatory, democratic</b>	<b>Participatory (democratic), consultative</b>	<b>Democratic and Free Schools</b>
<b>Collaborative, teamwork</b>	<b>Collaborative, teamwork</b>	<b>Cooperative Learning Teacher as facilitator, delegator</b>
<b>Self-directed groups</b>	<b>Holacracy, self-management in groups</b>	
<b>Laissez-faire, free-wheeling</b>	<b>Laissez-faire</b>	<b>Open Schools (and Classrooms) (Summerhill)</b>
<b>Chaotic</b>	<b>Chaotic</b>	
<b>People try their own thing</b>		<b>Inquiry-based learning, student-centered (related to constructivism) Teacher as facilitator, delegator</b>



# **Part 3**

## **General observations on knowledge organization and its role**

# 3.1 Better data modeling

- **Entity-relationship modeling is fundamental**

Kudos to Peter Chen (1976) and precursors

- **Three past blunders**

**1 Attributes as elements in entity-relationship modeling**

**2 Calling relationships *properties*, as is done in RDF**

**3 Using only binary (two-way) relationships**

# Part 4

# Conclusions

# Conclusions 1

- **Many applications of KOS.**
- **Consider both**
  - **requirements for machine processing, specifically inference, and**
  - **requirements for human processing, specifically meaningful arrangements that assists in making sense**

# Conclusions 2

- **Many opportunities for people with good training in KO to improve KOS now used**
- **Prepare students for that, specifically**
  - **Students should have a basic understanding of logic, formal ontology principles, inference, and complex queries**
  - **Foster the ability to discern meaningful structures and then convey structure and meaning through good document design.**
  - **Foster the ability to work with researchers on defining variables, determining data collection methods, and curating, and sharing data , all to improve interoperability and reusability.**

# Conclusions 3

- We need **more communication between the following largely separated communities:**
  - Knowledge Organization
  - Semantics in linguistics and terminology
  - Knowledge representation in artificial intelligence
  - Ontology
  - Data Modeling
  - Semantic Web

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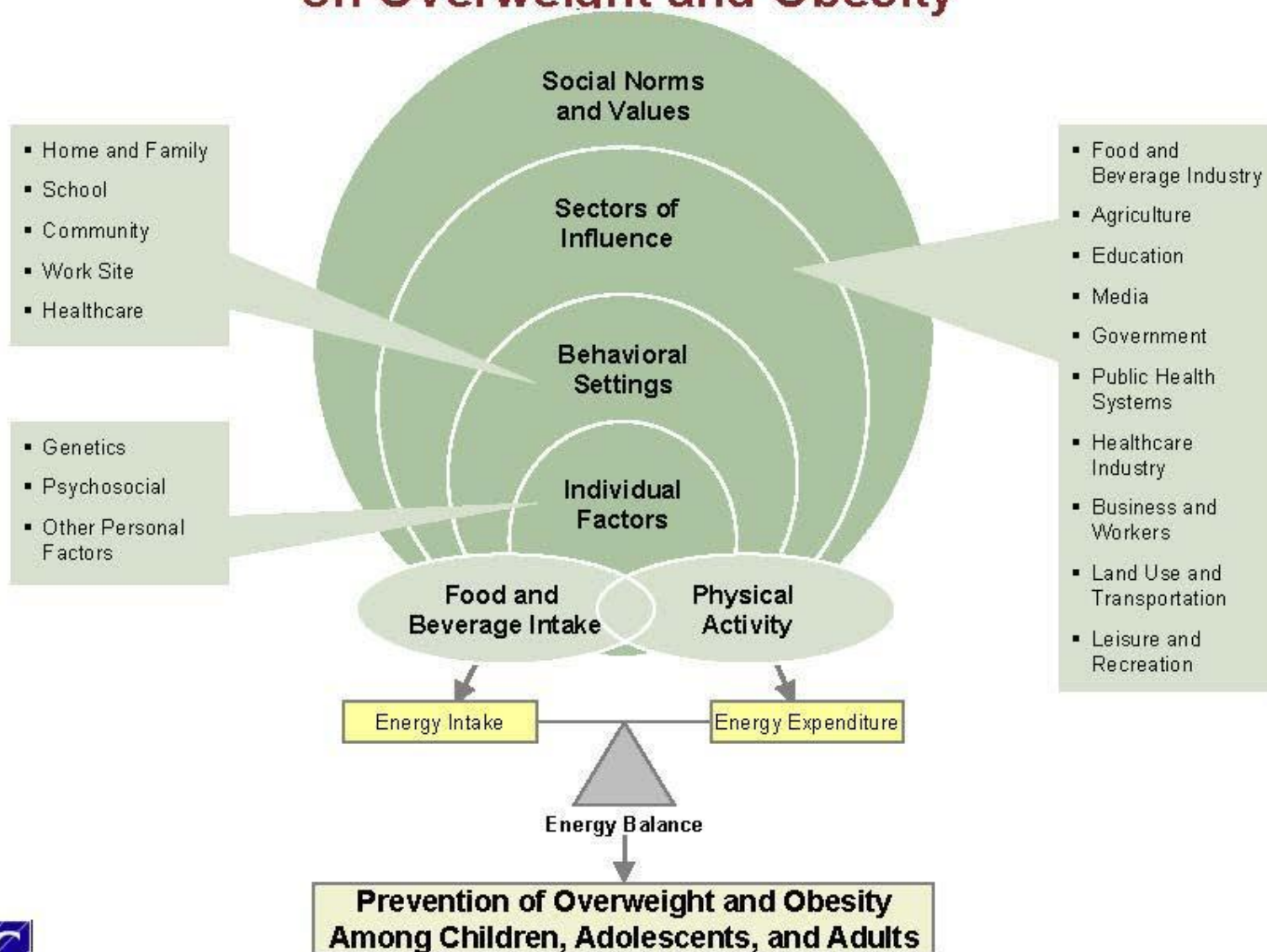
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**www.dsoergel.com**



# An Ecological Framework for Organizing Influences on Overweight and Obesity



Adapted from: Koplan JP, Liverman CT, Kraak VI, editors. Preventing childhood obesity: health in the balance. Washington, DC: Institute of Medicine, National Academies Press; 2005.

