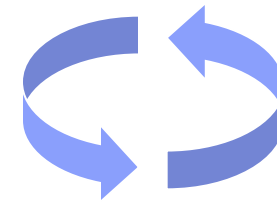


# Knowledge Analytics Synergy in Clinical Decision Support



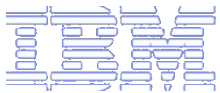
***MIE 2012***

*Noam Slonim, Boaz Carmeli, Abigail Goldstein, Oliver Keller, Carmel Kent, Ruty Rinott*

*IBM Haifa Research Labs*

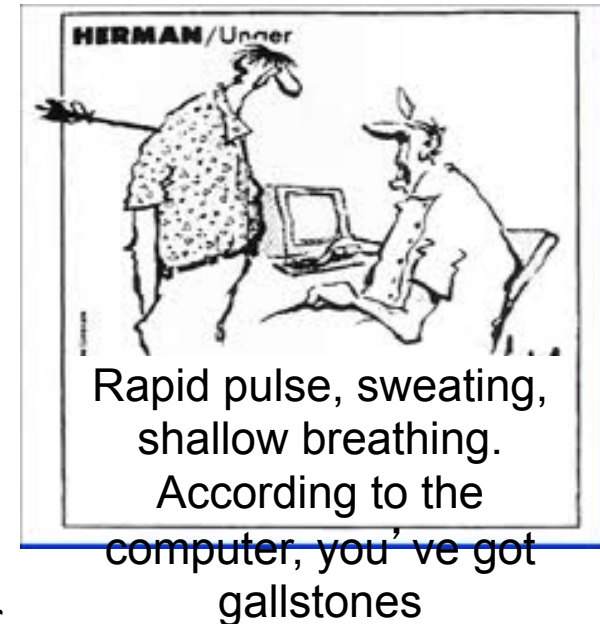
*Machine Learning and Data Mining (MLDM) group*

*Cli-G group*

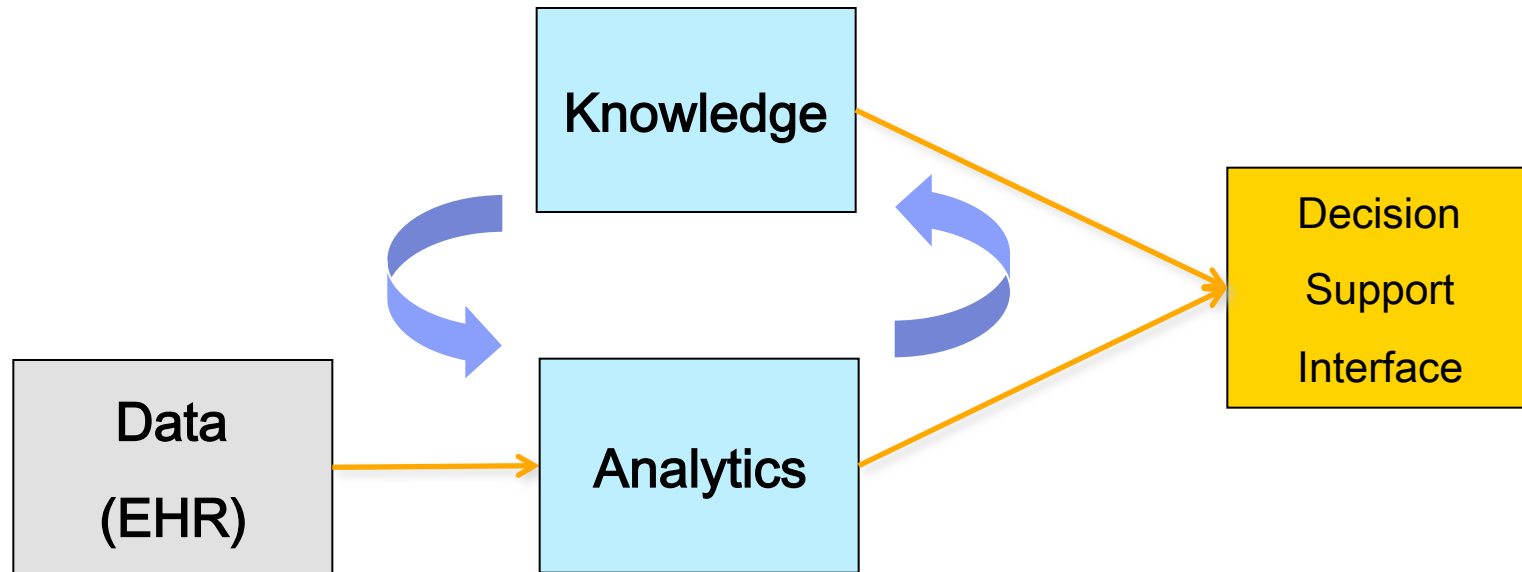


# Introduction

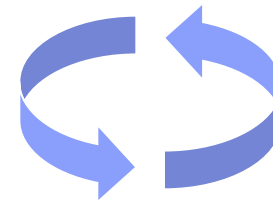
- Clinical Decision Support (CDS) have great potential
  - Improving clinical care
  - Reducing associated costs
- Most existing tools are knowledge-based CDS
  - Clinical trials → Clinical guidelines → simple CDS rules
  - Well established
- Emerging alternative paradigm – analytics-based CDS
  - EHR data → Analytics (Machine Learning) → Statistically-based CDS
  - No need to define/maintain long list of rules
  - Allows to take new “omic” data into account quickly
  - Address groups not participating in clinical trials (co-morbidities, elderly, etc.)



# Both paradigms are complementary



- Analytics techniques may benefit from domain knowledge
- Domain knowledge may benefit from analytics
- CDS knowledge-Analytics Synergy (KAS) paradigm



# The proposed CDS KAS Paradigm

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- Synergistically combine relevant clinical knowledge with analytics applied to EHR data, improving overall CDS quality
- Knowledge and analytics components mutually feed and enhance each other to achieve better results.
- Implemented as part of a generic CDS system – Cli-G
  - Cli-G underlying architecture explicitly supports the KAS paradigm
  - See also Evicase presentation (Wed. 1000, Fermi)

# Data examined to demonstrate KAS potential

- 3 cohorts of hypertensive patients:\*
- EPOGH
  - 1,149 patients
  - 15 clinical features
- IMA\_Sardinia
  - 278 patients
  - 45 clinical features
  - Genotype data for 1M SNPs
- Immidiet
  - 258 patients
  - 45 clinical features
  - Genotype data for 1M SNPs

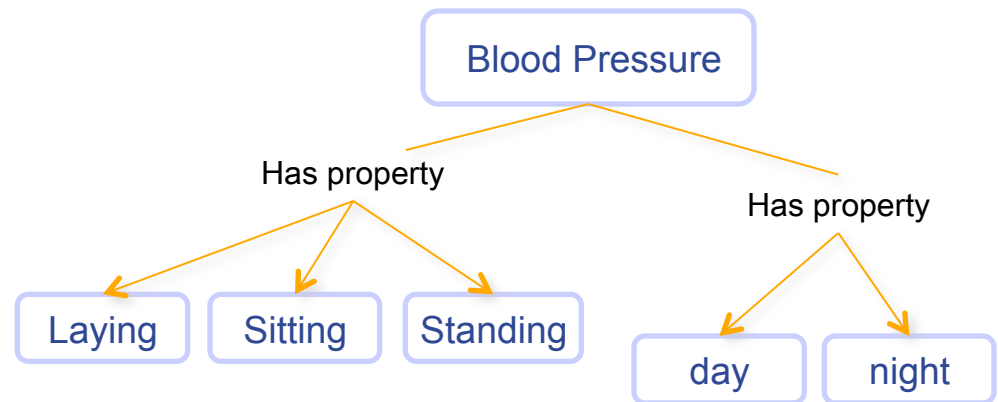


# The Knowledge-Management (KM) component

- Represents and streams relevant clinical knowledge to other components.

- Declarative Knowledge (DK)

- Factual information:  
hierarchy of concepts with  
properties and relations



- Procedural Knowledge (PK)

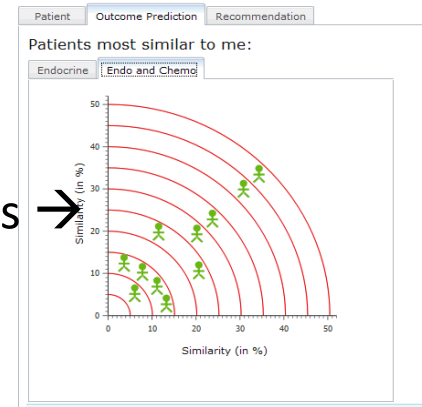
- The knowledge on how to operate upon the DK concepts
  - Clinical Guidelines
  - Rules defining Deductive Concepts to be added to the DK model; e.g.,
    - Risk Group based on Gender, Weight, Co-morbidities, etc

- May support various CDS applications

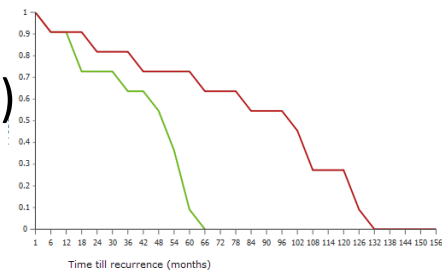
- Guidelines based treatment recommendations
- Rule-based alerts for adverse-drug events

# The Analytics component

- Encompasses an arsenal of ML algorithms for various CDS tasks
- Feature ranking/selection
  - Automatically highlight clinical features of potential interest to the clinician
- Estimating patient similarity
  - Automatically reveal meaningful groups of similar patients recommendations at Point Of Care
- Supervised Machine Learning algorithms
  - Predicting most common treatment (IHI 2012)
  - Predicting the outcome of candidate treatment (EuResist)
  - More...



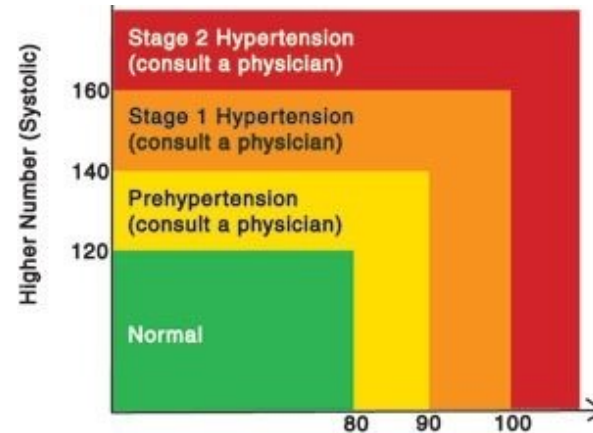
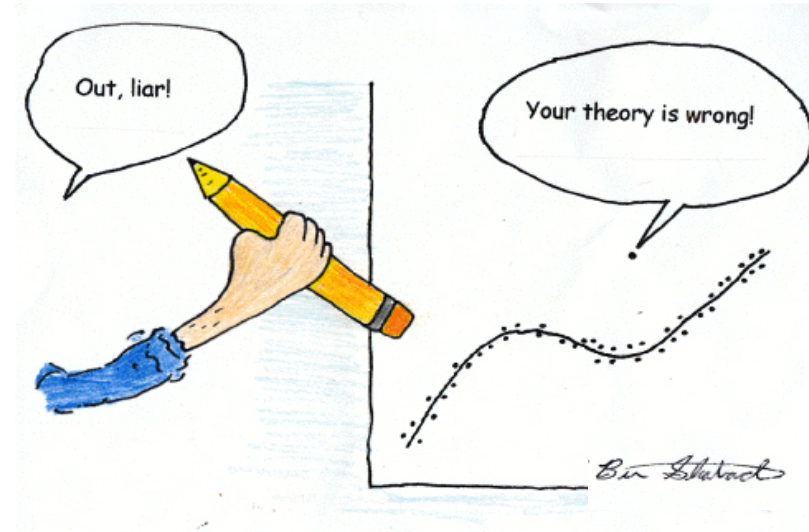
Recurrence Free Probability:



# KAS: using KM component to enhance Analytics

## ➤ Procedural Knowledge rules for **Data Cleansing**:

- Outliers / non-valid data removal
  - Remove Height measurements  $< 140$
  - Remove Serum Creatinine  $> 2$
- Unifying data reported in different scales
  - Serum Insulin:  $\mu\text{U/mL}$  or  $\text{pmol/L}$
- Quantizing continuous data
  - Systolic Blood Pressure:
    - $< 120 \rightarrow$  Normal
    - $120-139 \rightarrow$  Pre Hypertension
    - $140-159 \rightarrow$  Stage-1
    - $> 160 \rightarrow$  Stage-2





# KAS: using KM component to enhance Analytics

## ➤ Data enrichment

➤ Use domain knowledge to pinpoint relevant feature combinations to explore

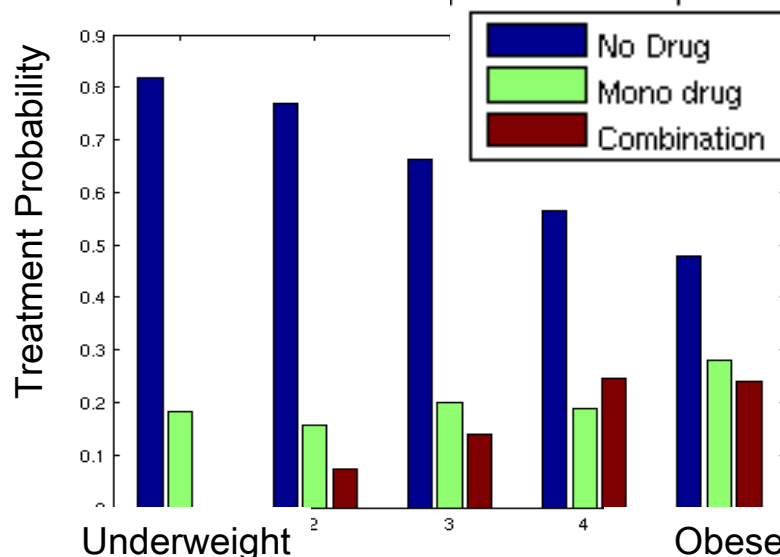
➤ BMI

➤  $[\text{Serum-Glucose} * \text{Serum-Insulin}] / 25 \rightarrow \text{HOMA1}$

➤ More...

➤ Some Deductive Concepts were later identified by feature-ranking as associated with clinicians treatment decision;

➤ BMI in the EPOGH cohort  $\rightarrow$  insights regarding treatment-selection process\*



\* Rinott *et. al*/IHI 2011

# KAS: using KM component to enhance Analytics

## ➤ Data filtering

- Use domain knowledge to filter irrelevant features, focusing analytics on important features
  - Find similar patients based on genetic SNP data
  - Using all 1M SNPs available
    - Time costly
    - Low signal to noise ratio
  - Pre-select 5,717 SNPs according to domain knowledge - potentially related to molecular pathways associated with hypertension
    - Reduce run time by more than 2 orders
    - Clustering patient based on similarity results in meaningful clusters

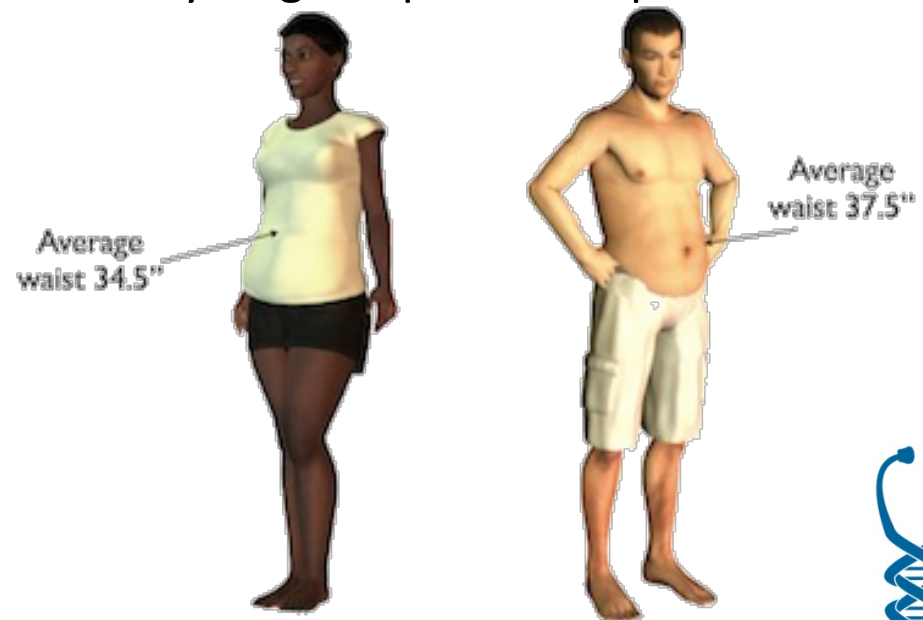
# KAS: using Analytics to enhance KM component

## ➤ Enhancing Declarative Model

- EHR often includes free-text data
- Analytics may reveal the importance of a particular free-text term; e.g.,
  - Drug-name → added to the DK Model

## ➤ Deductive Concepts

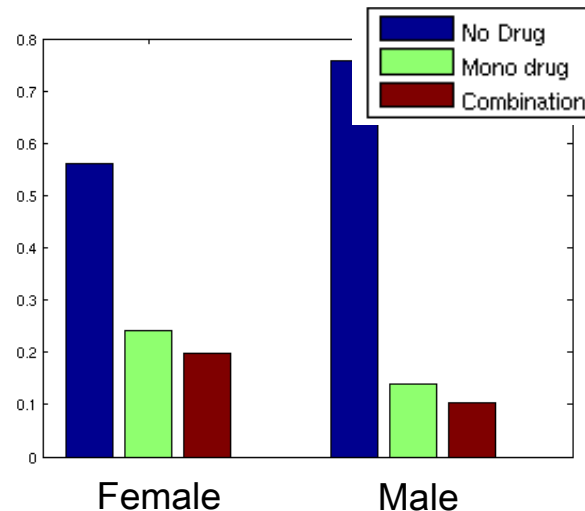
- Exploring all feature pairs → Reveal pairs with synergistic predictive power  
→ TBA as Deductive Concepts to the DK model
- Waist circumference & Gender has synergistic predictive power for selected treatment in EPOGH data



# KAS: using Analytics to enhance KM component

## ➤ Enhancing Procedural Knowledge

- Reveal clinicians “common practices” → TBA to the PK model → supporting rule-based CDS applications
- Feature ranking applied to identify features associated with treatment selection
- Gender found to be significantly associated with treatment selection
- Gender highlighted in PK model as related to “Common Practices” in “Treatment Selection”



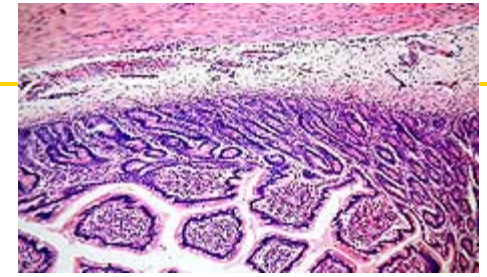
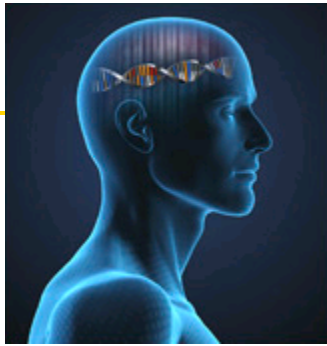
## ➤ Guidelines refinement

- Pinpoint SNPs of potential clinical value to enhance PK and DK → guidelines refinement (Not implemented yet)

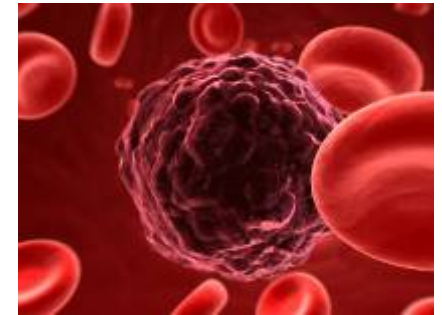
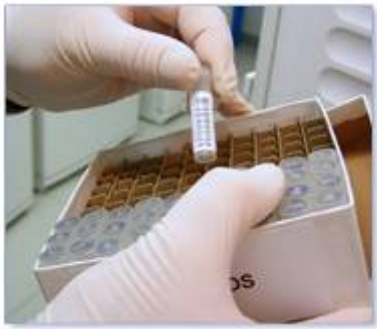
# Summary

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- Presented the Knowledge Analytics Synergy paradigm for CDS
  - Demonstrated via a generic tool (Cli-G) that relies on the KAS Architecture
- Presented concrete examples over clinical and genomic data:
  - Knowledge can enhance Analytics
  - Data-driven analytics can enhance Knowledge
- The synergistic approach holds great potential for CDS tools



**Thank You**

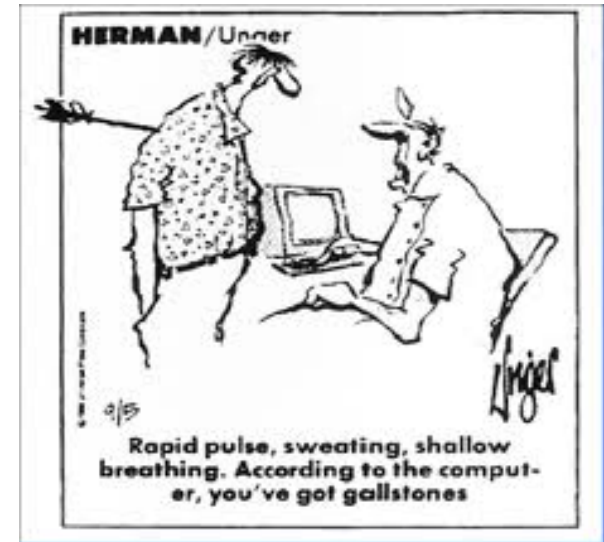


# Introduction

- Clinical Decision Support (CDS) have great potential
  - Improving clinical care
  - Reducing associated costs
- Most existing tools are knowledge-based CDS
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- Emerging alternative paradigm – analytics-based CDS
  - EHR data → Analytics (Machine Learning) → Statistically-based CDS rules



Clinical Genomics IT



# Results – Analytics enhancement – **Data Filtering**

- Patients were clustered using Patient Similarity defined according to SNP data
  - Emphasizing relative similarity of patients sharing identical rare SNPs
- Similarity estimated from 5,717 SNPs (from 1M) selected by domain knowledge
  - Filtering reduced computation time by more than two orders of magnitude
  - Filtering focused analysis in advance on potentially most relevant SNPs
- Statistical significance of clinical features was estimated in each cluster
  - c10 (30 people)
    - Avg. Systolic BP: 153, vs. 140 in remaining population (P-val 0.004)
  - c5 (23 people)
    - Avg. Triglyceride: 95.3, vs. 132 in remaining population (P-val 0.0002)
  - More...
- → Encouraging results that should be further explored ...



# Both paradigms are potentially complementary

- Analytics-based Insights extracted from EHR data may better reflect the relevant population
- Analytics-based CDS require no need to define/maintain long list of rules
- Inherently stochastic statistical signals may better reflect clinical world than deterministic rules
- Being blind to the relevant clinical knowledge is obviously hazardous
  - Contraindications
- Analytics techniques may benefit from domain knowledge

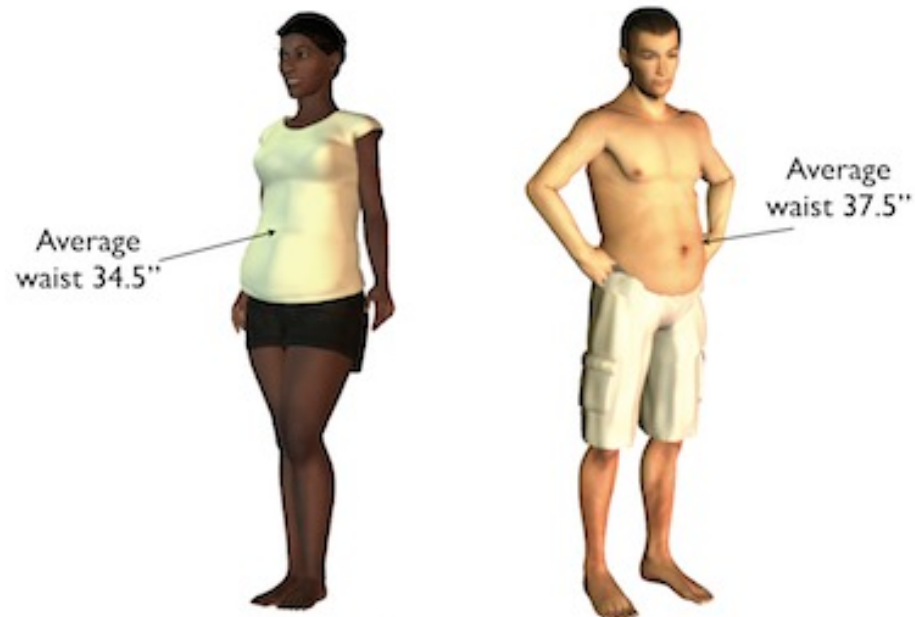


CDS knowledge-Analytics Synergy (KAS) paradigm



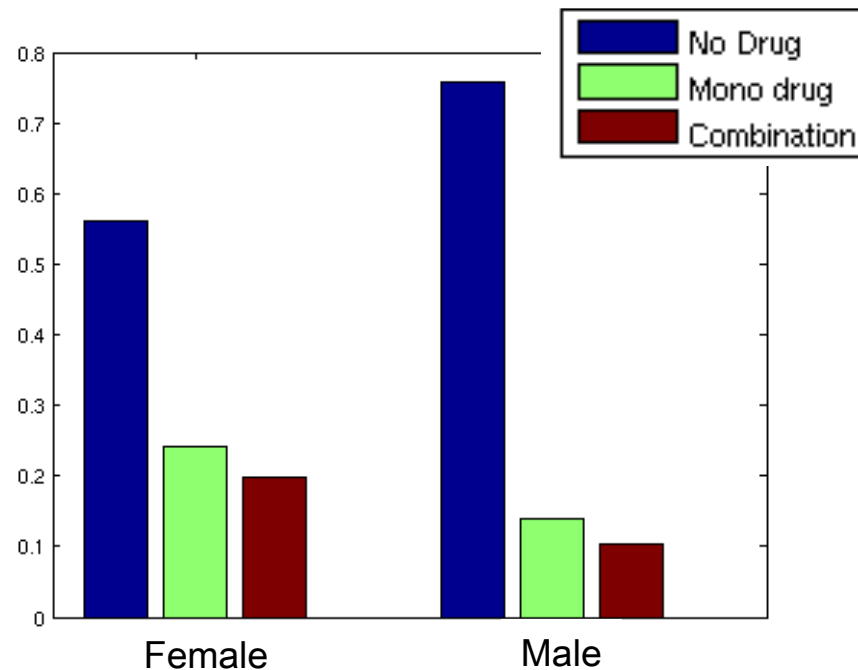
# Results – Knowledge enhancement

- Searched for feature-pairs with synergistic predictive power for selected treatment in EPOGH data
  - Several pairs were detected as significantly associated with treatment allocation:
    - Waist circumference & Gender → added to DK model as Deductive Concepts



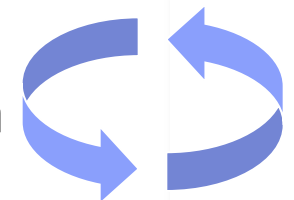
# Results – Knowledge enhancement

- Feature ranking applied to identify features associated with treatment allocation
  - Gender found to be significantly associated with treatment selection
  - → Gender highlighted in PK model as
    - related to “Common Practices” in “Treatment Selection”
  - Relevant reference exist
    - → Accompany analytics-based evidence with pointer to relevant literature...?



# Both paradigms are complementary

- Being blind to the relevant clinical knowledge is obviously hazardous
  - Contraindications
  - Rare conditions
- Analytics techniques may benefit from domain knowledge
- Analytics-based insights may highlight important features not taken into account by current guidelines
- Analytics-based Insights extracted from EHR data may better reflect population
  - Address groups not participating in clinical trials (co-morbidities, elderly, etc.)
  - Consider new genetic information
- Domain knowledge may benefit from analytics



CDS knowledge-Analytics Synergy (KAS) paradigm

# KAS CDS High level architecture

