

# Agent based Monitoring of Gestational Diabetes Mellitus

René Schumann, Stefano Bromuri, Johannes Krampf,  
Michael I. Schumacher

[rene.schumann@hevs.ch](mailto:rene.schumann@hevs.ch), [stefano.bromuri@hevs.ch](mailto:stefano.bromuri@hevs.ch), [michael.schumacher@hevs.ch](mailto:michael.schumacher@hevs.ch)

## 1. Our Motivation: Gestational Diabetes

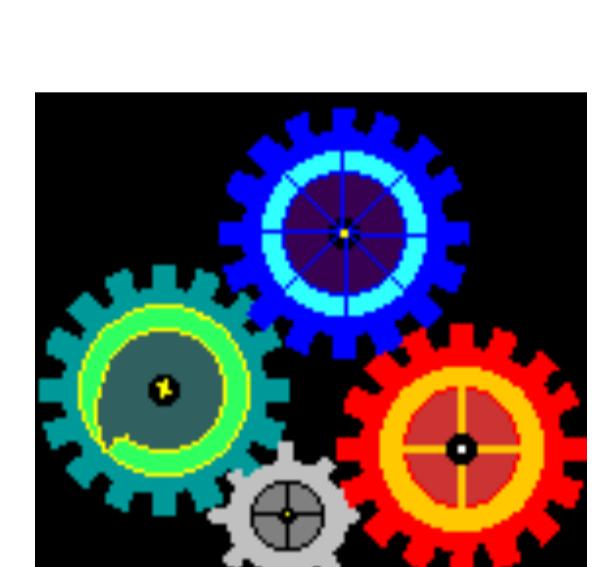
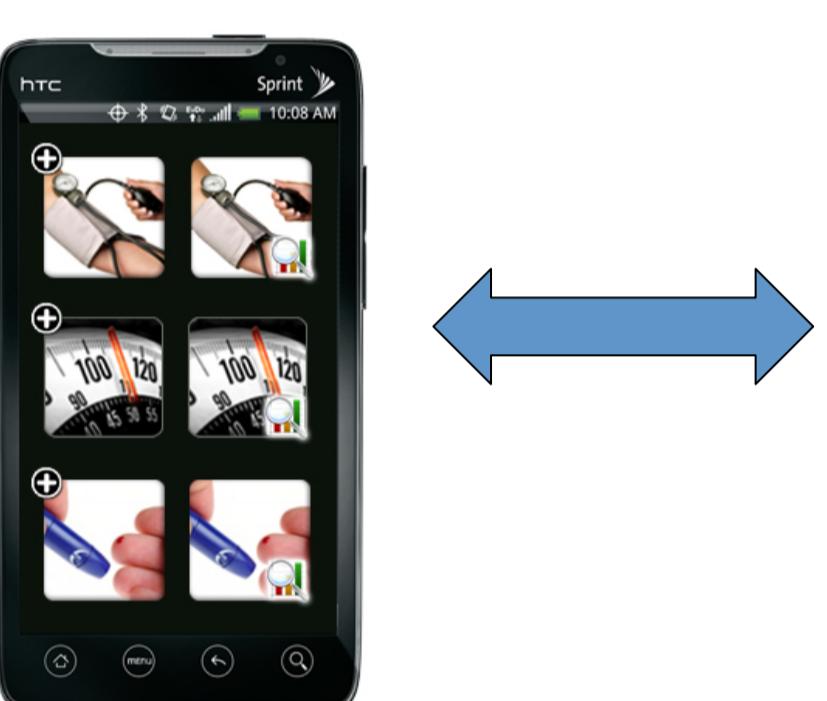


- Gestational diabetes mellitus (GDM) is a condition that affects 2-5% of all the pregnancies and manifests itself with high blood sugar levels during pregnancy.
- It can cause health problems for mother and child like,
  - Preeclampsia, Eclampsia,
  - Hyperglycemia,
  - Macrosomia, and
  - Diabetes type II
- Current treatment practices:  
Doctors usually see patient 2-3 times a week. In case of hyperglycemia, the patient may arrive **too late**.



## 2. Our Goals: Improve the Level of Care

- Goal of our system
  - **Improve** the care level of the woman
  - **React faster** to the development of possible life-threatening conditions.



- Means to do so:
  - **Pervasive Health Systems** for acquiring data
  - **Multi-Agent System** for reasoning about the data



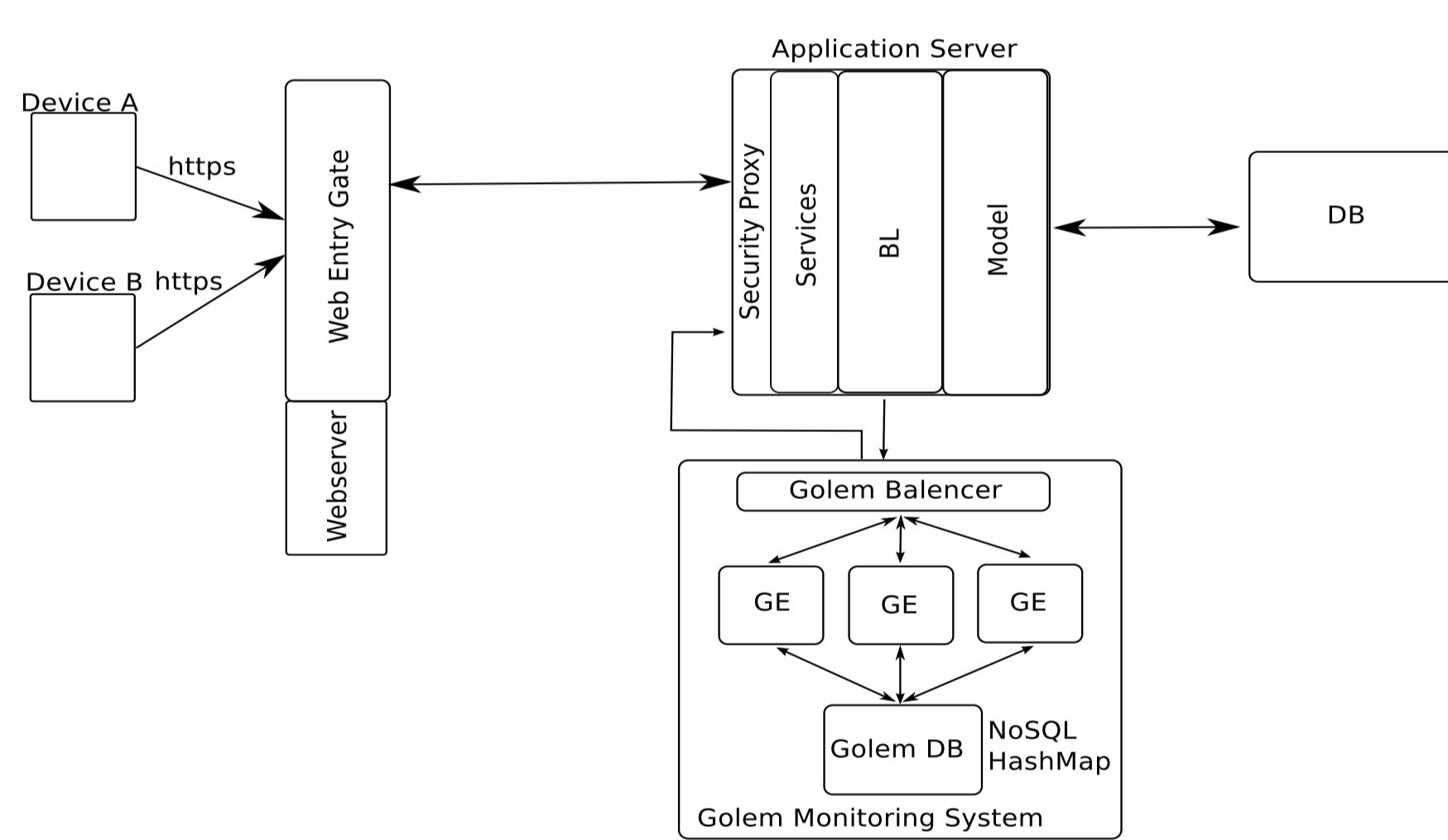
## 3. The Pervasive Healthcare System

- Patients can enter their physiological values, using a mobile App developed in the project
- The data is send to the server when the phone has a data connection, otherwise it is stored until a connection has been established



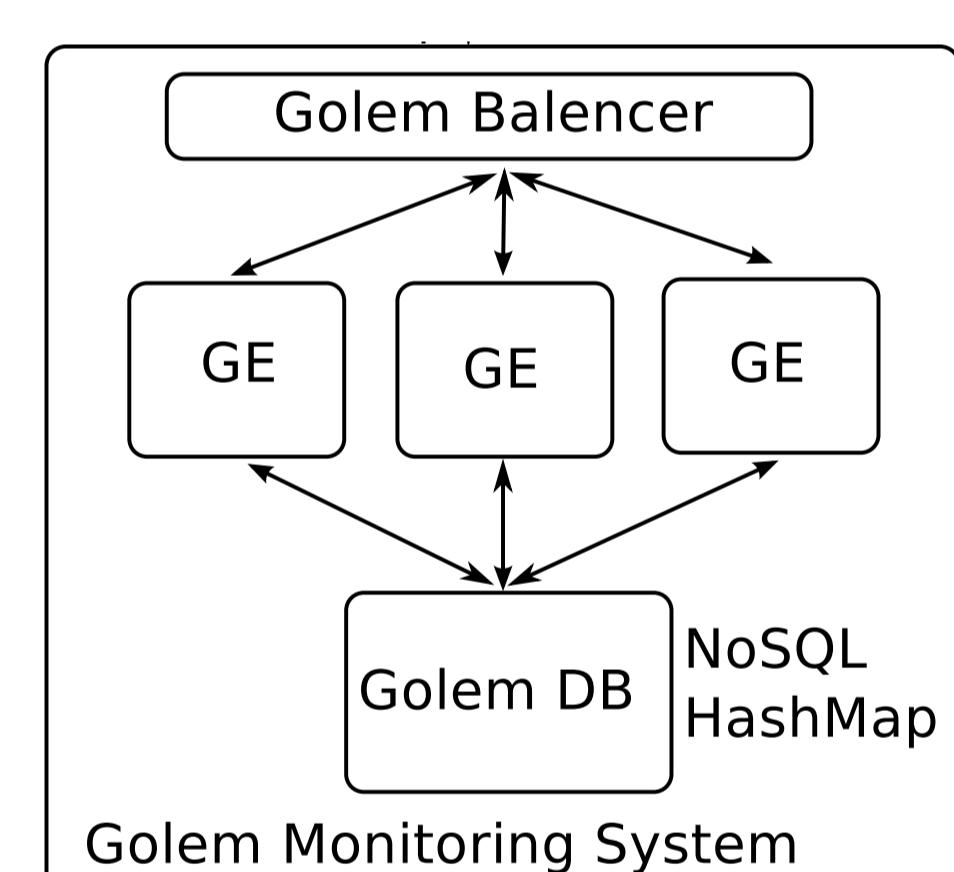
- Incoming patient data is
  - Stored in a database
  - Forwarded to an agent-based analyzing component

### Architecture of the Service Infrastructure



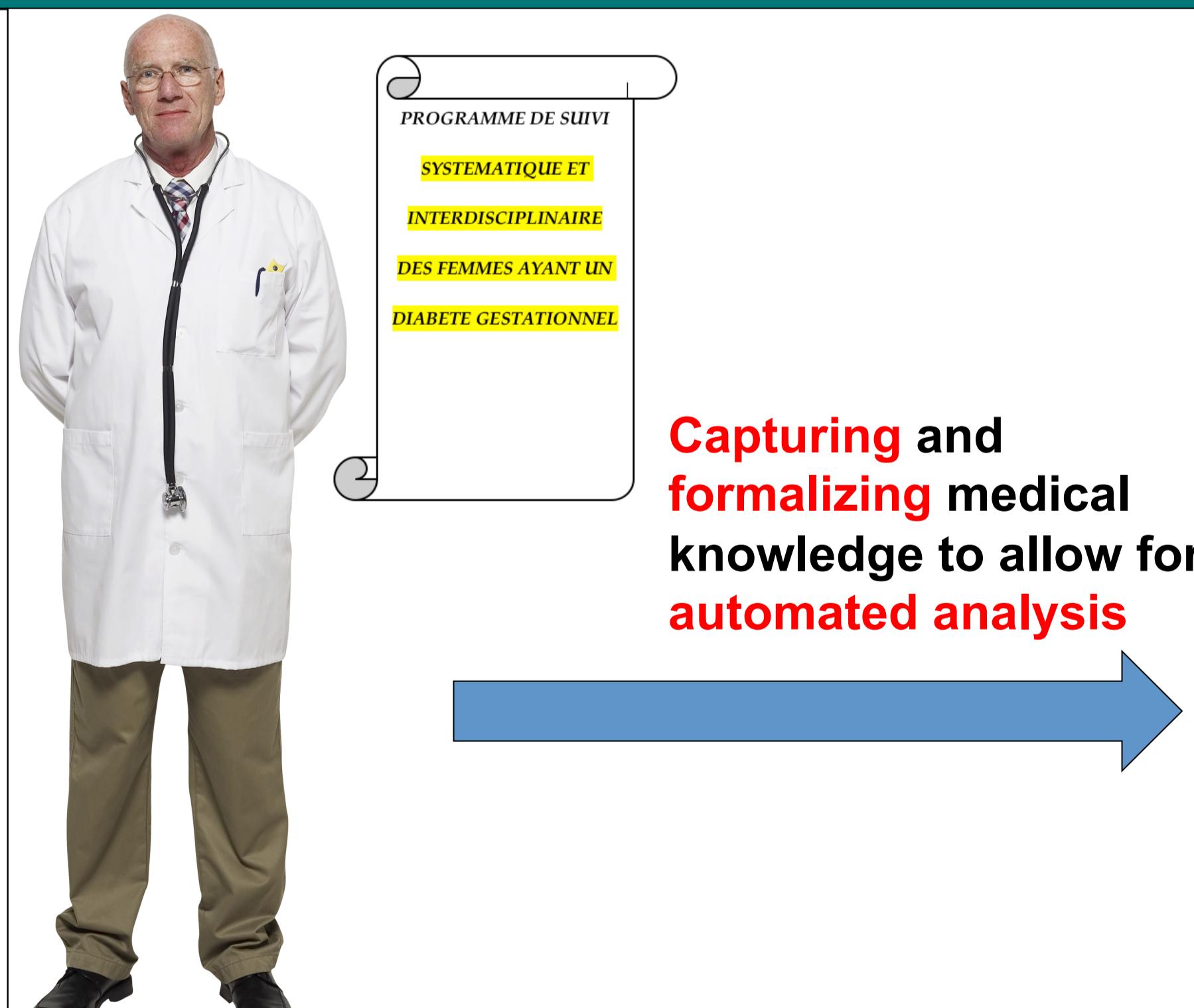
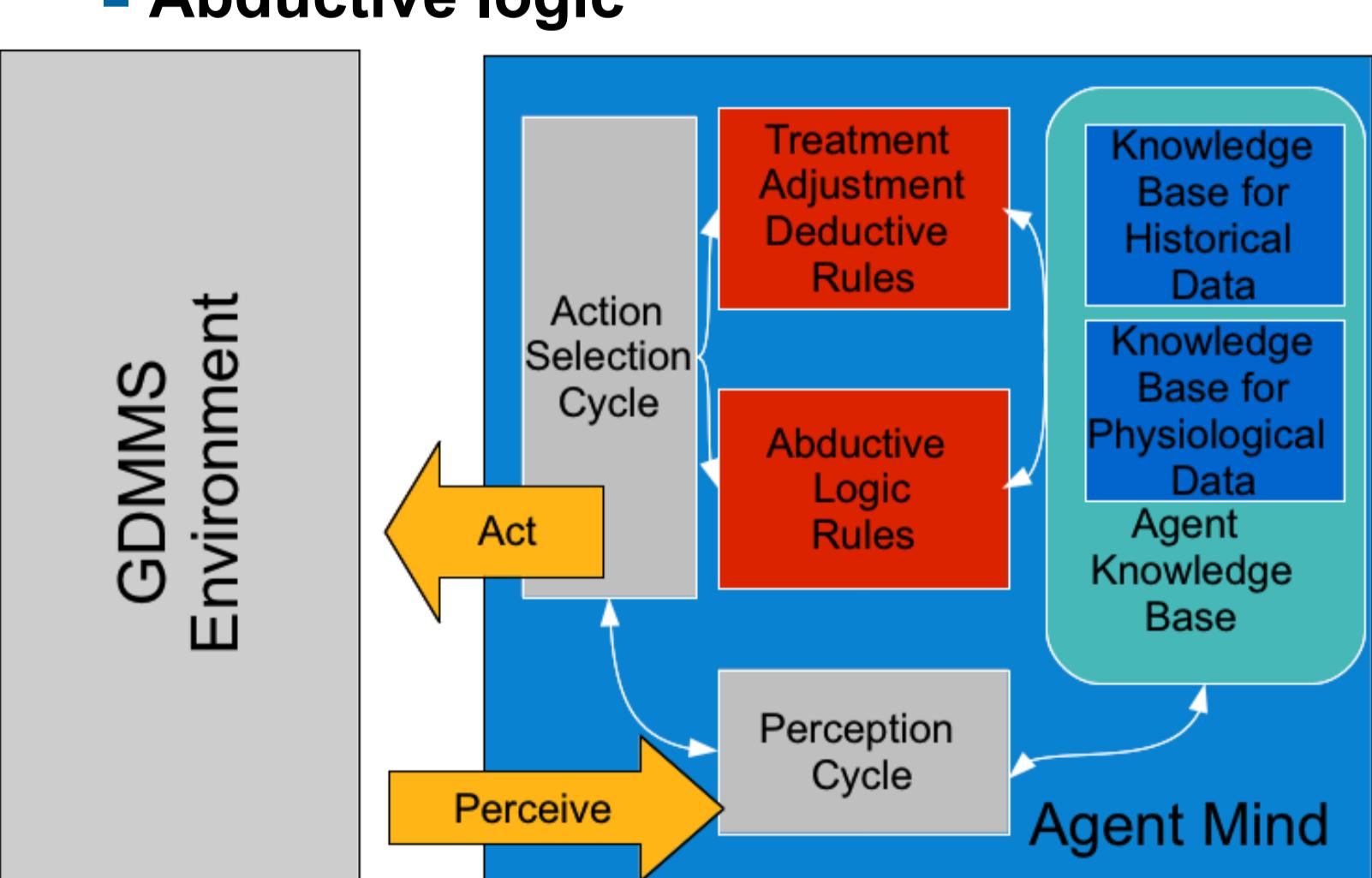
## 4. The Multiagent System

- An agent is responsible for monitoring one patient
- To scale-up, the agents can be:
  - distributed on different nodes;
  - activated and deactivated by a balancer;
  - persisted in an agent database.



## 4. The Multiagent System (continuation)

- Each agent has capabilities to efficiently reason about recognized Events.
- Medical conditions can be declared in terms of
  - Deductive logic
  - Abductive logic



### Abductive Rules

Domain Knowledge :  
previous\_gestational\_week = macrosomia.

week\_of\_gestation(W) ← W >= 34; macrosomia; W >= 20; preeclampsia.  
glucose\_observation(A) = [Gl, Gl<sub>1</sub>, ..., Gl<sub>n</sub>] ←  
  Gl, Gl<sub>j</sub> ∈ A, j ≠ i, Gl<sub>i</sub> > 5, Gl<sub>j</sub> > 5; macrosomia.  
previous\_preeclampsia\_blurred\_vision; severe\_headache;  
low\_blood\_pressure; proteinuria; preeclampsia; severe\_preeclampsia.  
kidney\_disease; oliguria; proteinuria; nausea; epigastric\_pain ←  
preeclampsia; severe\_preeclampsia.  
blood\_pressure(Systolic, Diastolic) ←  
preeclampsia; proteinuria, Systolic >= 140, Diastolic >= 90;  
not proteinuria; severe\_hypertension, Systolic >= 160, Diastolic >= 110;  
Systolic >= 140, Diastolic >= 90; proteinuria, week\_of\_gestation(W), W < 35.  
(severe\_headache) ∨ (urine\_proteinuria); severe\_preeclampsia.  
blood\_pressure(Systolic, Diastolic) ←  
preeclampsia, (not proteinuria); week\_of\_gestation(W), W < 20;  
not proteinuria; severe\_hypertension.  
preeclampsia; blood\_pressure(S.D.S) < 140, D < 90.  
IC :  
← preeclampsia, (not proteinuria); week\_of\_gestation(W), W < 20;  
not proteinuria; severe\_hypertension.  
select(Symptoms, A, T) ←  
  demo(Symptoms, Explanation).instance\_of(D.patient,T).myID(D.ID),  
  member(preeclampsia), Explanation).full\_piers(D.ID,Probability,T);  
  A = store\_event;C(factor->AD.physiological\_data#2!type=observation,  
  timestamp = T, value => Probability;preeclampsia].

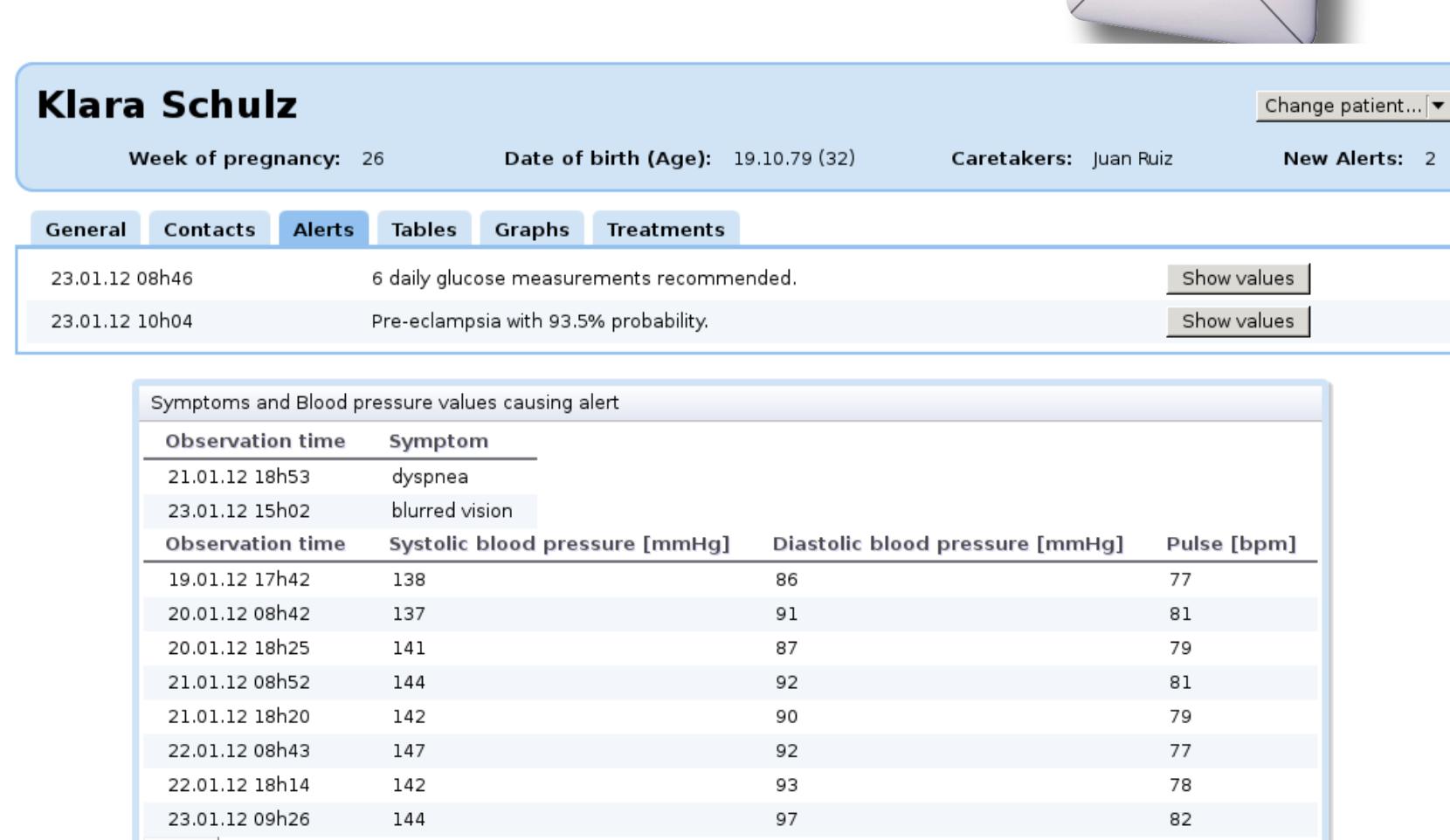
full\_piers(D,Probability,T) ← holds\_att(D,platelet,PLA,T), holds\_att(D,spo2,spo2,T),  
holds\_att(D,creatinine, CRE, T), holds\_att(D,aspartate, AST, T),  
holds\_att(D,gestational\_age, Age,T), holds\_att(D,cholesterol, Chol, T),  
Coef<sup>1</sup> = 1.68 + (1.54 \* Age) - 1.23 \* (Age<sup>2</sup>) + (2.71 \* 10<sup>-2</sup> \* CRE) +  
(2.07 \* 10<sup>-3</sup> \* PLAs) + (4.10 \* 10<sup>-5</sup> \* PLAs<sup>2</sup>) + (1.04 \* 10<sup>-4</sup> \* AST) +  
(-3.05 \* 10<sup>-6</sup> \* AST<sup>2</sup>) + (2.50 \* 10<sup>-5</sup> \* CRE\*PLA) + (-6.99 \* 10<sup>-5</sup> \* PLAs\*AST) +  
(2.56 \* 10<sup>-3</sup> \* PLA\*spo2), Probability = exp(Coef)/1+exp(Coef).

### Deductive Rules

R1) terminates(glucose\_observation;Ev, ID, observations, \_).  
R2) initiates(glucose\_observation;Ev;ID,observations,[Ev|Tail]) ←  
  patient\_of(Ev, ID), time(Ev, T), holds\_att(ID,observations, Tail, T).  
R3) initiates(glucose\_observation;Ev;ID,hypoglycemia\_risk, true) ←  
  patient\_of(Ev, ID), time(Ev, T), holds\_att(ID,observations, List, T),  
  size(List, 20), not(member(M, List), glycemia(M, GI), GI > 4).  
R4) initiates(glucose\_observation;Ev, ID, hypoglycemia\_risk, true) ←  
  patient\_of(Ev, ID), glycemia(E, GI), time(Ev, T),  
  holds\_att(ID,observations, [Head|Tail]), time(Head,T'),  
  T - T' < 1, glycemia(Head, GI2), GI < 3, GI2 < 3.  
R5) initiates(glucose\_observation;Ev;ID,postprandial\_observation,[Ev|Tail]) ←  
  is\_postprandial(Ev), time(Ev, T), patient\_of(Ev, ID),  
  holds\_att(ID,postprandial\_observation, Tail, T).  
R6) initiates(glucose\_observation;Ev;ID,introduce\_preprandial,true) ←  
  holds\_att(ID,postprandial\_observation, List, T),  
  sublist(Sub,List), size(Sub,6), foreach(Obs,Sub,(glycemia(Obs,G),G>8)).  
select(A,T) ← instance\_of(P, patient,T), holds\_att(P,introduce\_slow\_insulin,true,T),  
  myID(D), A = store\_event;C(factor->D,recommendation,p[  
  timestamp => T, value => introduce\_slow\_insulin]).

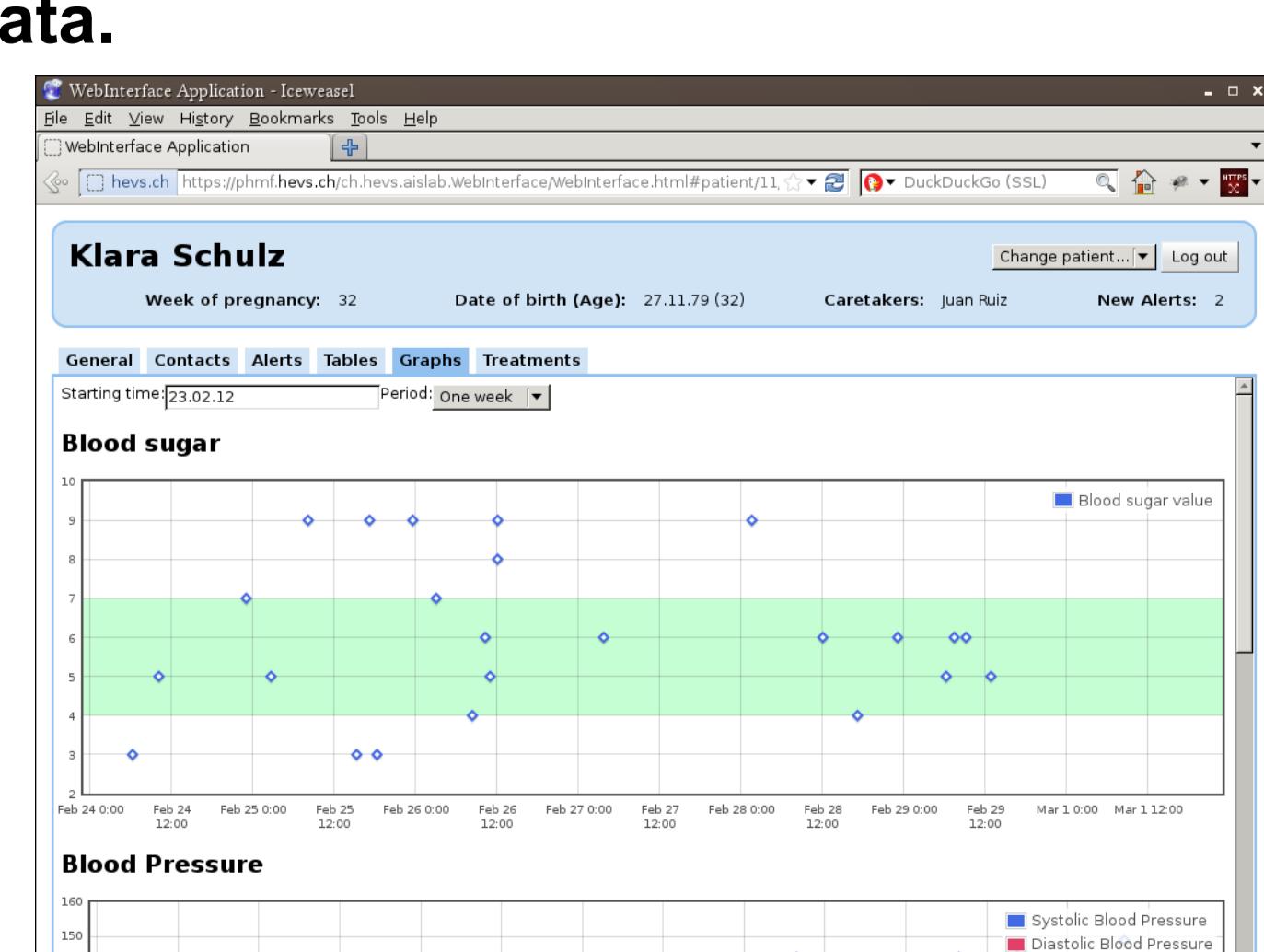
## 5. The Web Interface

- If an alert occurs:
  - The doctor in charge gets notified via E-Mail



- He can see all details at the Web interface

- Graphs and tables help to access patients data.



## 6. Status of the project

- Field tests will start in the second half of 2012



- Patients will be given a smart phone to monitor their condition for three month each.

