
Relevance for Pharma Research



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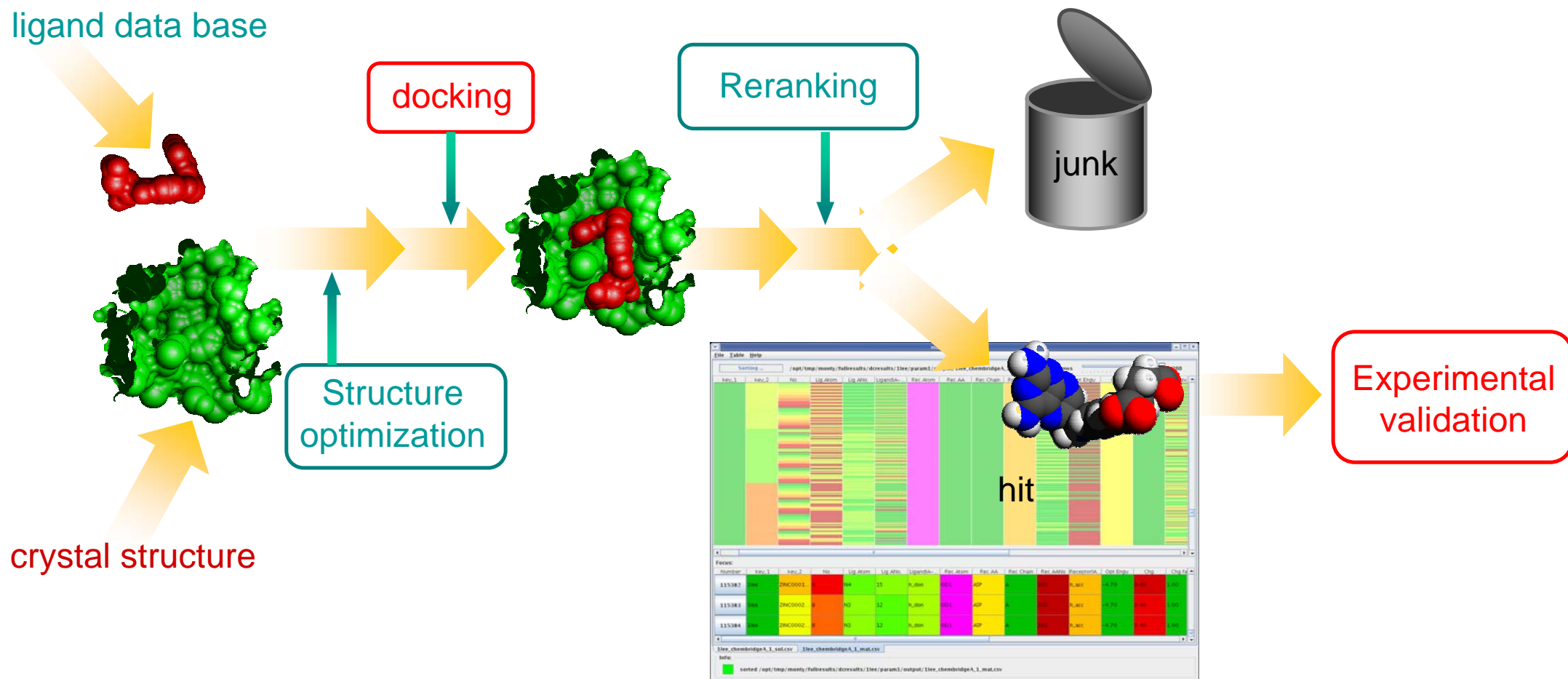
Fraunhofer Institute for Algorithms and Scientific Computing (SCAI)
and

Bonn-Aachen International Center for Information Technology (B-IT) /
Friedrich-Wilhelms-University of Bonn

GRID Computing for Pharma R&D

- Distributed Virtual Screening – the WISDOM Example
 - Basics of Structure-based Virtual Screening
 - Docking on the GRID: WISDOM and follow-up
- Text Mining on the GRID – Information Extraction for Scientific & Competitive Intelligence
 - Scientific & Competitive Intelligence
 - Distributed information extraction from scientific literature

Basics of Structure-based Virtual Screening



Structure – based Virtual Screening on the GRID

- **Not** a new idea
- Simple task farming approach possible
- Routine procedure at Novartis and other pharma front runners
- In EnterpriseGRID – solutions typically based on proprietary middleware platforms (e.g. United Devices (UD); Plattform Computing)
- Success stories available e.g. identification of cyclin dependent kinase inhibitors published by Novartis
- Close interaction between *in silico* and “wet” laboratory world required

Docking on the GRID: WISDOM and follow-up

WISDOM : Wide In Silico Docking On Malaria

Biological goal

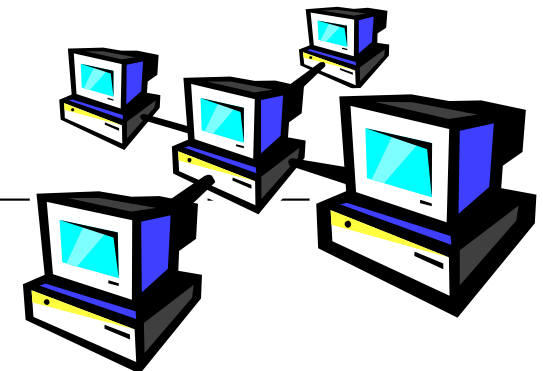
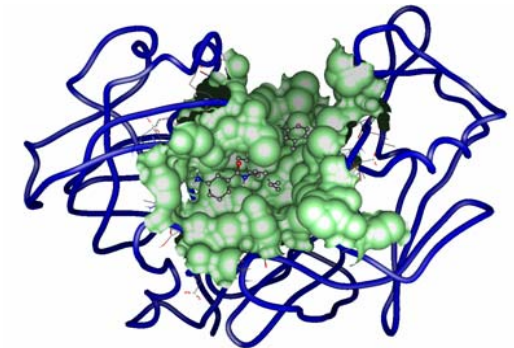
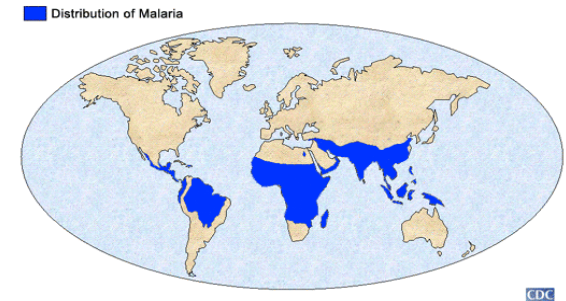
Proposition of new inhibitors for a family of proteins produced by *Plasmodium falciparum*

Biomedical informatics goal

Deployment of *in silico* virtual screening on the grid

Grid goal

Deployment of a CPU consuming application generating large data flows to test the grid operation and services
→ “data challenge”



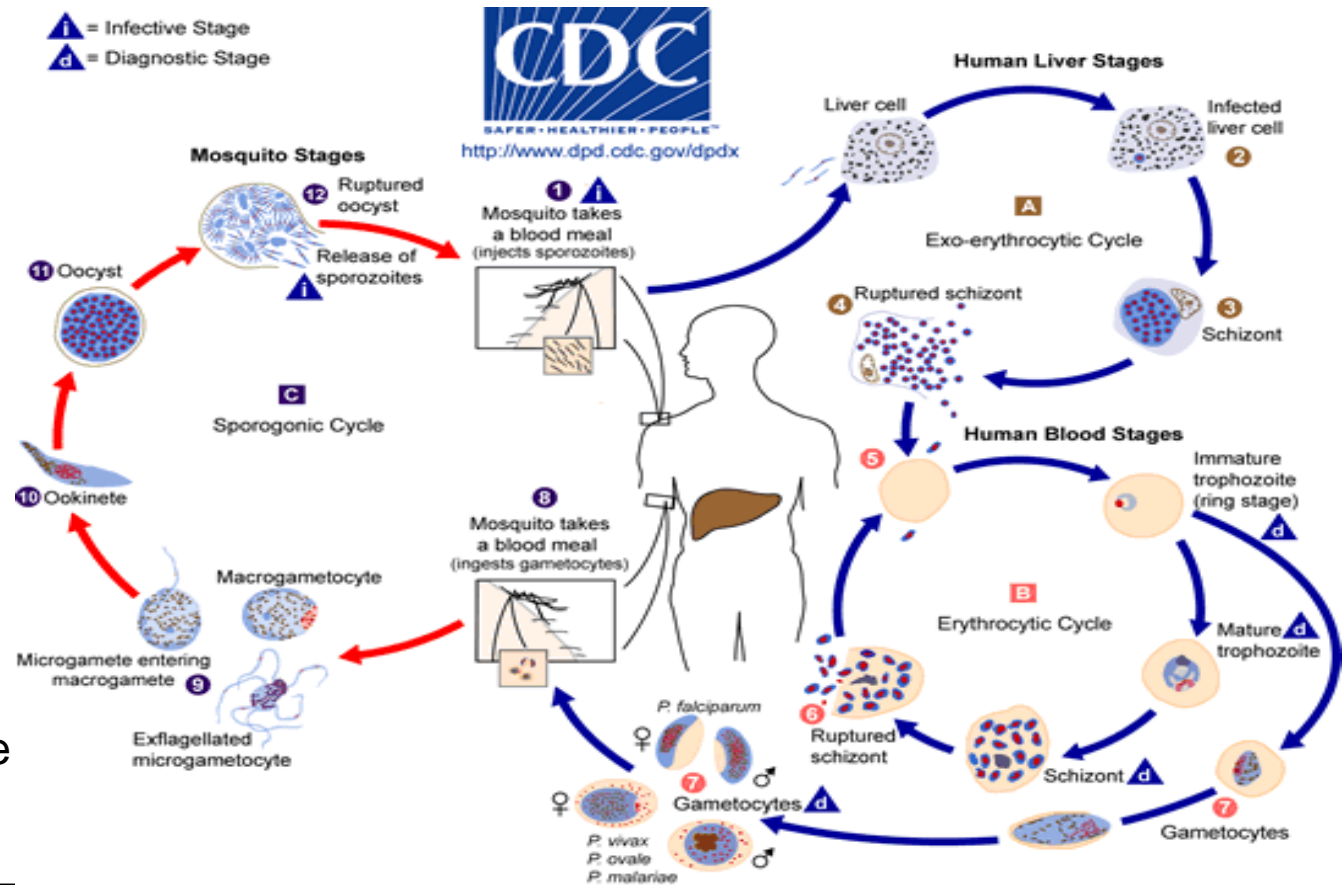
Introduction to the Disease : Malaria

~300 million people worldwide are affected

1-1.5 million people die every year

Widely spread

Caused by protozoan parasites of the genus *Plasmodium*



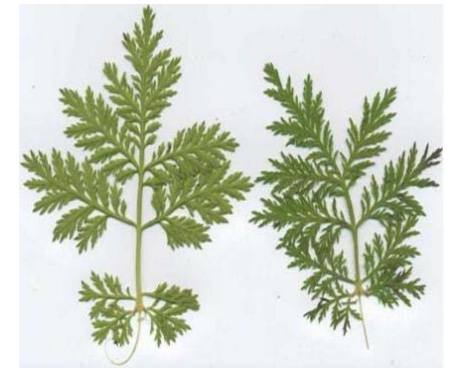
Strong Need for New Drugs against Malaria (WHO)

Drug resistance has emerged for all classes of antimalarials except artemisinins.

- ❑ Resistance to chloroquine, the cheapest and the most widely used drug, is spreading in almost all the endemic countries.
- ❑ Resistance to the combination of sulfadoxine-pyrimethamine which was already present in South America and in South-East Asia is now emerging in East Africa

All countries that experience resistance to conventional monotherapies should use ACTs (artemisinin-based combination therapies)

But there is even the threat of resistance to artemisinin too, as it is already observed in murine *Plasmodium yoelii*



Identification of New *Plasmodium* Targets

There is consensus that substantial scientific effort is needed to identify new targets for anti-malaria drugs

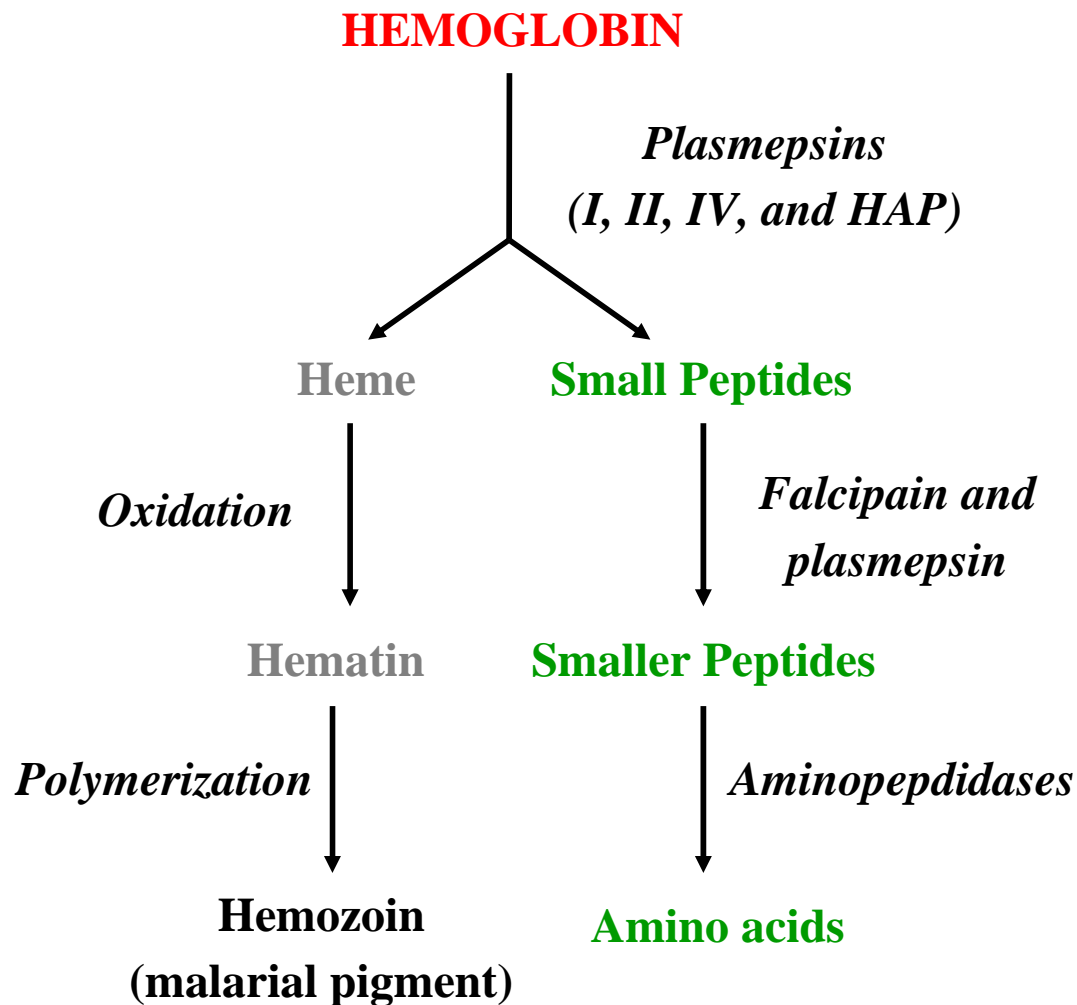
With the advent of the *Plasmodium* genome, many targets came into light

The potential anti-malarial drug targets are broadly classified into three categories, and each category has many individual targets.

- ❑ Targets involved in human hemoglobin degradation (proteases)
- ❑ Targets involved in parasite metabolism (Folate, phospholipid...)
- ❑ Targets engaged in parasite membrane transport and signalling (choline carrier etc).

WISDOM focuses on hemoglobin metabolism and especially on Plasmepsin II and Plasmepsin IV

Plasmepsins and Their Role in Human Hemoglobin Degradation



Plasmepsins are involved in hemoglobin degradation inside the food vacuole during the erythrocytic phase of the parasite life cycle.

Plasmepsins are present in all of the four species of *Plasmodium* causing the disease in human

Sequence homology between the different plasmepsins is high (65-70%)

Sequence homology with its nearest human aspartic protease neighbour is fortunately low (35%)

Crystallographic data of plasmepsins are available in PDB

EGEE, the World's Largest Grid Infrastructure

Started in 2004, +70 partners in the world

Project leader : CERN

6 scientific domains with >20 applications deployed

170 grid nodes, 17000 CPUs, several PetaBytes of data, 10000 jobs by day

BioMed VO

27 Computing Elements (~3.000 CPUs)

28 Storage Elements (~21 TB disks)

in 12 countries



Countries with nodes supporting the data challenge WISDOM

VS Explorer: a Tool for Analyzing “Grid Scale” Ranking Lists

The image displays two windows of the VS Explorer software, illustrating its functionality for analyzing grid-scale ranking lists.

Left Window (mainWindow): Shows a large grid of 400 rows. The columns are labeled: Number, SMILES, name, scenario1, and scenario2. A red oval highlights a specific row, which is magnified in the right window.

Right Window (mainWindow): Shows a detailed view of the selected row (row 28) and its corresponding chemical structure and ranking data across 10 scenarios.

Row 28 Data:

Number	SMILES	name	scenario1	scenario2	scenario3	scenario4	scenario5	scenario6	scenario7	scenario8	scenario9	scenario10
25		ZINC00603011	-28.92	-29.88	-28.66	-28.08	-27.14	-28.66	-28.08	-28.91	-28.92	-29.88
26		ZINC00605829	-19.20	-17.29	-19.49	-24.32	-20.74	-19.49	-24.32	-19.20	-18.66	-17.29
27		ZINC00606383	-9.60	-8.35	-10.59	-12.48	-10.59	-10.45	-12.19	-10.45	-10.45	-8.35
28		ZINC00607811	+00.01	+00.01	+00.01	+00.01	+00.01	+00.01	+00.01	+00.01	+00.01	+00.01

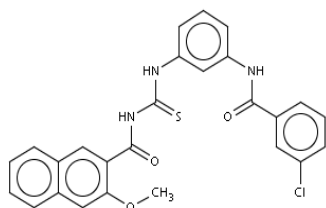
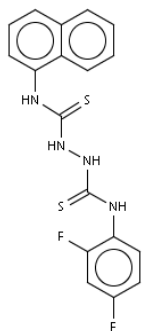
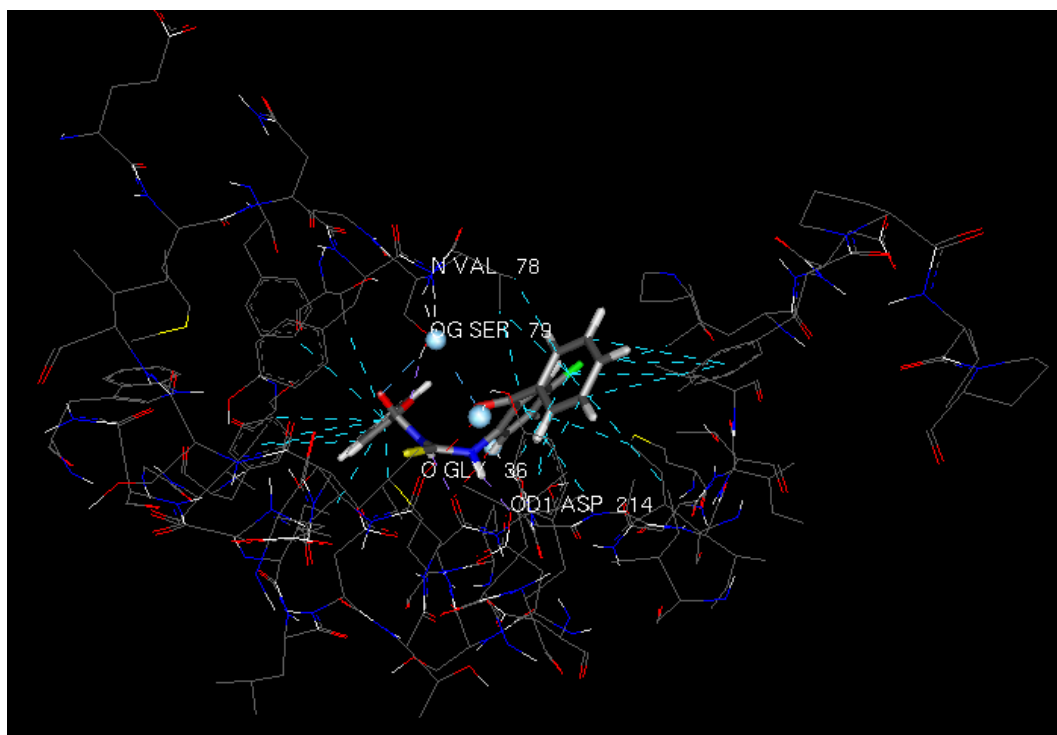
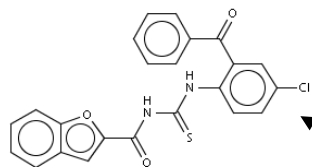
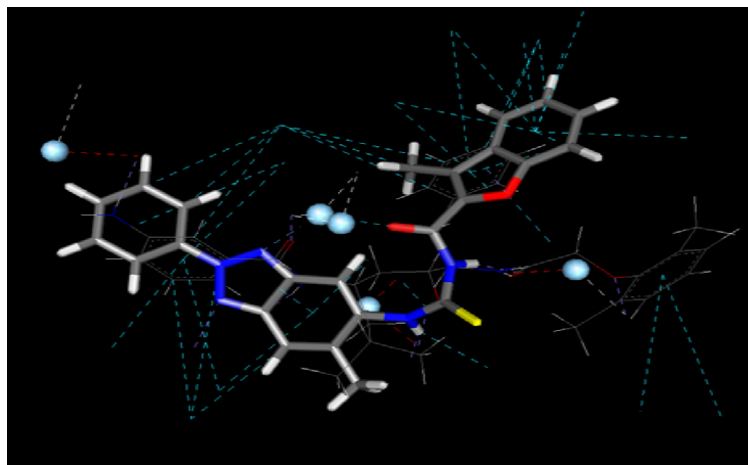
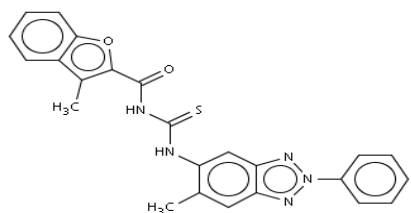
Focus:

Number	SMILES	name	scenario1	scenario2	scenario3
62		ZINC0062...	+00.01	-14.24	+00.01
63		ZINC0062...	+00.01	-15.52	+00.01
64		ZINC0062...	-35.64	-37.31	-37.16

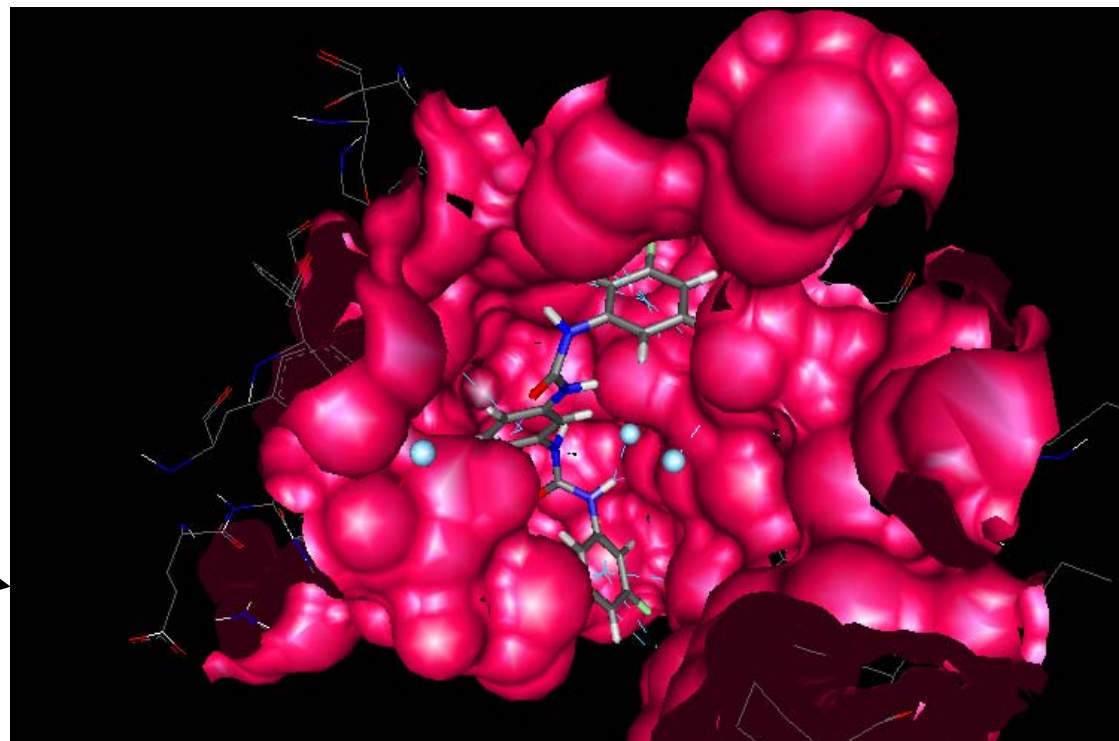
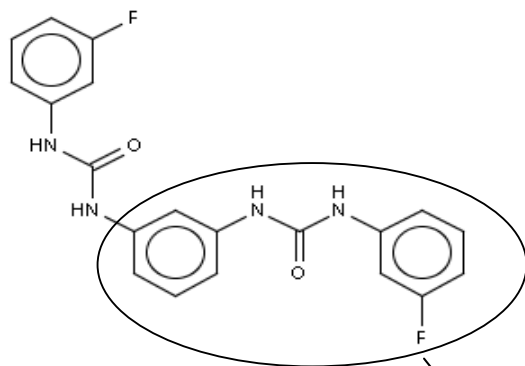
Info:

loaded /home/bio/groupshare/dcp/param.csv

Compounds for MD - Thiourea compounds



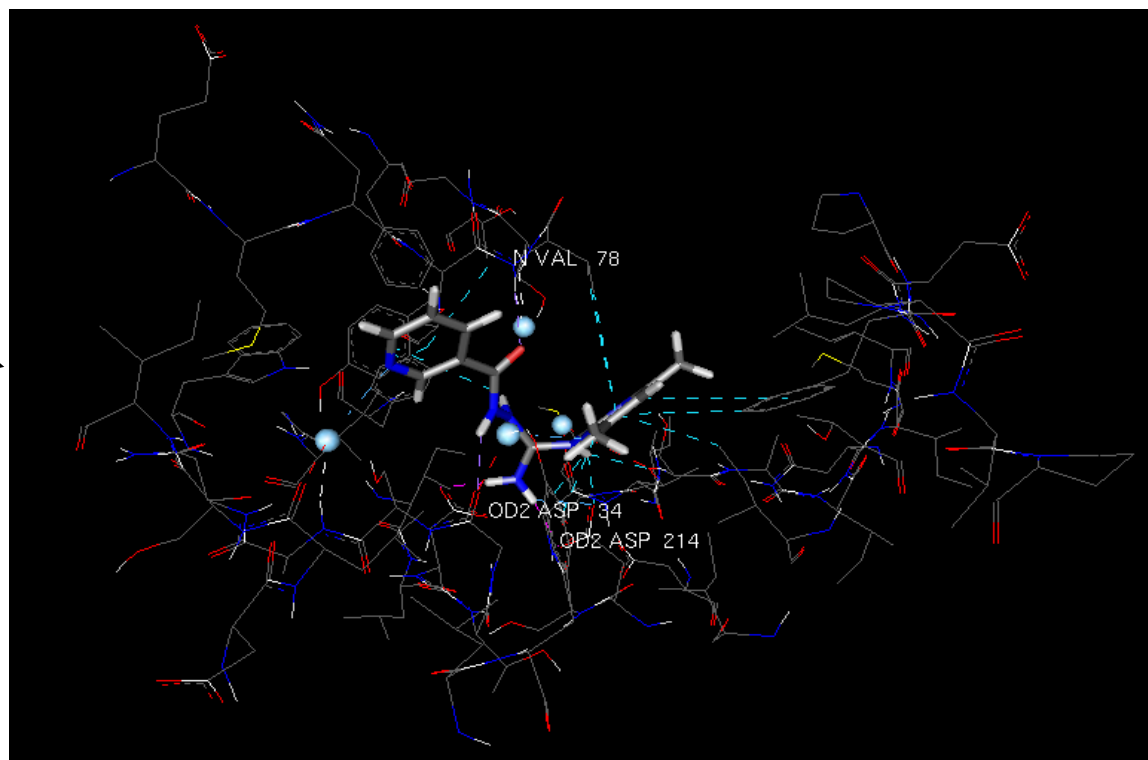
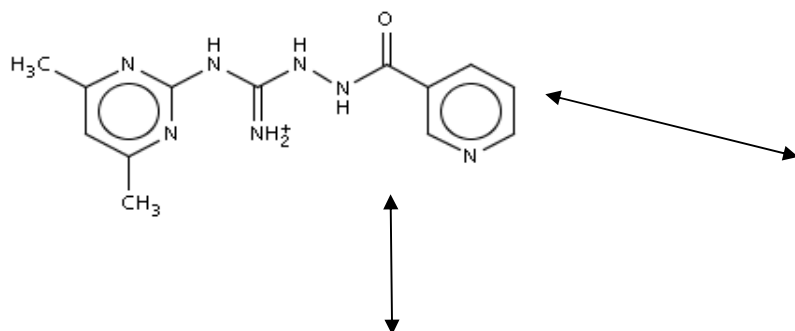
Compounds for MD-Urea compounds



Note: Diphenyl urea compounds are well in agreement with literature (Walter Reed compounds)

Terminal								
File	Edit	View	Terminal	Tab	Help			
No.	Lig.	Lig.	Ligand	Rec.	Rec.	Rec.	Rec.	Receptor
Atom	ANo.	IA-Type	Atom	AA	Chain	AANo	IA-Type	
1 N4	21 h_don	water				120	h_acc	
1 C18	25 phenyl_ring	CG	PHE	A		294	phenyl_center	
1 C15	22 phenyl_center	CE1	PHE	A		294	phenyl_ring	
1 C15	22 phenyl_center	CG2	VAL	A		78	ch3_phe	
1 C8	11 phenyl_center	C	THR	A		217	amide	
1 C8	11 phenyl_center	C	GLY	A		216	amide	
1 C8	11 phenyl_center	CD1	ILE	A		32	ch3_phe	
1 C8	11 phenyl_center	CG2	ILE	A		32	ch3_phe	
1 C8	11 phenyl_center	CE	MET	A		15	ch3_phe	
1 O1	9 h_acc	OG	SER	A		79	h_don	
1 N1	7 h_don	O	GLY	A		216	h_acc	
1 C2	2 phenyl_ring	CG	TYR	A		77	phenyl_center	
1 C1	1 phenyl_center	CD1	ILE	A		123	ch3_phe	
1 C1	1 phenyl_center	CD2	TYR	A		77	phenyl_ring	
1 C1	1 phenyl_ring	CG	TYR	A		77	phenyl_center	
1 N3	18 h_don	OD2	ASP	A		34	h_acc	
1 N3	18 h_don	OD1	ASP	A		34	h_acc	
1 C15	22 phenyl_center	CE2	TYR	A		192	phenyl_ring	
1 C15	22 phenyl_center	CG1	VAL	A		78	ch3_phe	
1 N4	21 h_don	OD1	ASP	A		214	h_acc	
1 C20	27 phenyl_ring	CG	TYR	A		192	phenyl_center	
1 C15	22 phenyl_center	CD1	ILE	A		300	ch3_phe	

Compounds for MD- Guanidino compounds



Terminal								
File Edit View Terminal Tabs Help								
No.	Lig.	Lig.	Ligand	Rec.	Rec.	Rec.	Rec.	Receptor
	Atom	A No.	IA-Type	Atom	AA	Chain	A A No.	IA-Type
1	N1	5	h_acc	water			58	h_don
1	N7	19	h_acc	water			39	h_don
1	N7	19	phenyl_center	C	TYR	A	77	amide
1	C7	13	amide	CG	TYR	A	77	phenyl_center
1	C13	21	ch3_phe	CG	TYR	A	192	phenyl_center
1	N1	5	phenyl_center	CE1	PHE	A	294	phenyl_ring
1	N1	5	phenyl_center	CG2	VAL	A	78	ch3_phe
1	N1	5	phenyl_center	CD1	ILE	A	300	ch3_phe
1	N1	5	phenyl_center	CE2	TYR	A	192	phenyl_ring
1	N1	5	phenyl_center	CG1	VAL	A	78	ch3_phe
1	C3	3	phenyl_ring	CG	PHE	A	294	phenyl_center
1	N3	8	h_don	OG1	THR	A	217	h_acc
1	N3	8	h_don	OD1	ASP	A	214	h_acc
1	N4	10	h_don	OD1	ASP	A	34	h_acc
1	N4	10	h_don	OD2	ASP	A	214	h_acc
1	N4	10	h_don	OD1	ASP	A	214	h_acc
1	C12	20	phenyl_ring	CG	TYR	A	77	phenyl_center
1	N7	19	phenyl_center	CD2	TYR	A	77	phenyl_ring
1	O1	14	h_acc	N	VAL	A	78	h_don
1	N6	12	h_don	O	GLY	A	36	h_acc

Note: Satisfied all criteria, good binding mode, interactions to key residues, good score, appropriate descriptors.

Conclusions

- Virtual Screening is a straightforward approach to use the GRID in the pharma context
- Virtual Screening has been successfully used in EnterpriseGRIDs in the pharma industry and recently also on a large eScience infrastructure, the EGEE GRID
- Novel, promising candidate structures for the development of new anti-malarial drugs have been identified using GRID-based virtual screening
- WISDOM has initialized a series of follow-up projects that address other diseases such as avian bird flue
- The relevance of virtual screening approaches on the GRID has been proven; uptake by small and medium size pharma companies is still too slow

Text – Mining on the GRID – Information Extraction for Scientific & Competitive Intelligence

Scientific & Competitive Intelligence

What is Scientific & Competitive Intelligence ?

- Scientific and competitive intelligence are terms coined for the application of automated information mining methods
- Information mining ranges from improved document retrieval to full blown information extraction
- Goal of the pharmaceutical industry is to make sure that **all** relevant information is at hand when a decision about a drug development project is to be made
- Consequently, scientific and competitive intelligence encompasses not only text mining in PubMed abstracts, but extends to patent literature and business news streams

Protein Name Recognition

Multiple names for one gene

Ambiguous names in databases

Ambiguous acronyms

Common word names

Multi-word terms

Spelling variants

Permutations

Nested protein names

F12A	Neuronectin, GMEM, tenascin, HXB, cytotactin, hexabrachion
	p21, EPO, large T antigen
	WAS, STEP, iCE, StAR
	Interleukin 1 alpha Tumor necrosis factor beta
COL1A1	Collagen, type I, alpha 1 Collagen alpha 1(I) chain Alpha 1 collagen Alpha-1 type I collagen
	TNF receptor 1 collagen, type I, alpha receptor

Dynamically Labelled Text:

PMID- 11210439
OWN - NLM
STAT- medline
DA - 20010208
DCOM- 20010405
LR - 20031114
VI - 23
IP - 1
DP - 2001 Jan
TI - Fluid loading in rats increases serum **brain natriuretic peptide** concentration.
PG - 93-5
AB - Hyponatremia after subarachnoid hemorrhage has been linked to high plasma concentration of **atrial natriuretic peptide** and **brain natriuretic peptide**. Volume expansion therapy to prevent symptomatic vasospasm, such as intensive hypertensive and hypervolemic therapy, increases **atrial natriuretic peptide** concentration of these peptides. We therefore examined **brain natriuretic peptide** secretion in rats in response to acute volume expansion, infusing to 10 ml of saline over 1 h. In the 10 ml group, **brain natriuretic peptide** concentrations showed a significant increase from pre-infusion concentrations 1 h after initiation of infusion, but had begun to fall 1 h later. We suspect that high plasma concentration of **brain natriuretic peptide** after subarachnoid hemorrhage is partly caused by hypervolemic therapy.
AD - Department of Neurosurgery, Graduate School of Medical Sciences, Kyushu University, 3-1-1 Maidashi, Higashi-ku, 812-8582 Fukuoka, Japan. s-inoha@ns.med
FAU - Inoha, S
AU - Inoha S

Chemical Name Recognition

Dictionary names:

Brand names

Ariven, Extren, Clivarin

Organic chemical compounds

2-Acetoxybenzoic acid

Generic names, INN, USAN

Aspirin, Celecoxib, Heparin

Substance classes

secondary amine, cholin sulfates

Side groups, atoms and ions

butyl group, potassium, fluoride
anion

Pharmacological and biological effects

Cyclooxygenase inhibitor,
Anticoagulants

Regular expressions:

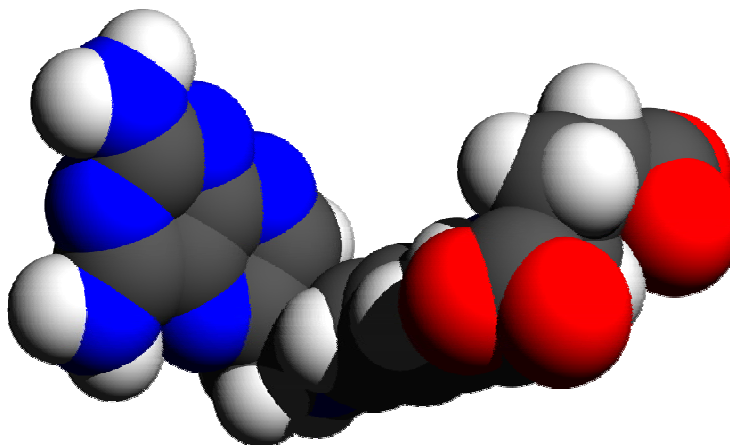
IUPAC names

N-[2-[4-[(2-oxy cyclohexyl)methyl]-

...

Representations of Chemical Compounds

- ☐ Name (trivial, trade, brand, INN, USAN)
- ☐ Registration numbers (CAS, NCI, Beilstein)
- ☐ Formal description (sum formula, SMILES)
- ☐ Chemical nomenclature (IUPAC, CAS, InChI)
- ☐ **Depictions**



Chemical Structure Recognition – an Overview

1 Document



US 20050182053A1

(19) **United States**

(12) **Patent Application Publication** (10) Pub. No.: US 2005/0182053 A1
Chen et al. (43) Pub. Date: Aug. 18, 2005

(54) SUBSTITUTED
3-AMINO-THIENO[2,3-B]PYRIDINE-2-
CARBOXYLIC ACID AMIDE COMPOUNDS
AND PROCESSES FOR PREPARING AND
THEIR USES

Publication Classification

(51) Int. Cl.⁷ A61K 31/5377; A61K 31/496;
A61K 31/4743
(52) U.S. Cl. 514/232.5; 514/301; 514/253.04;
544/125; 544/362; 546/114

(75) Inventors: **Zhidong Chen**, New Milford, CT (US); **Pier Francesco Cirillo**, Woodbury, CT (US); **Darren DiSalvo**, New Milford, CT (US); **Welmín Liu**, Sandy Hook, CT (US); **Daniel Richard Marshall**, Sandy Hook, CT (US); **Lifen Wu**, New Milford, CT (US); **Erick Richard Roush Young**, Danbury, CT (US)

(57) **ABSTRACT**
Disclosed are compounds of formula (1):

NC(=O)c1nc2c(ncn2c1R1)N3CCCN(C3)CC(R2)C(R3)Z

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(73) Assignee: **Boehringer Ingelheim Pharmaceuticals, Inc.**, Ridgefield, CT

(21) Appl. No.: 11/002,828

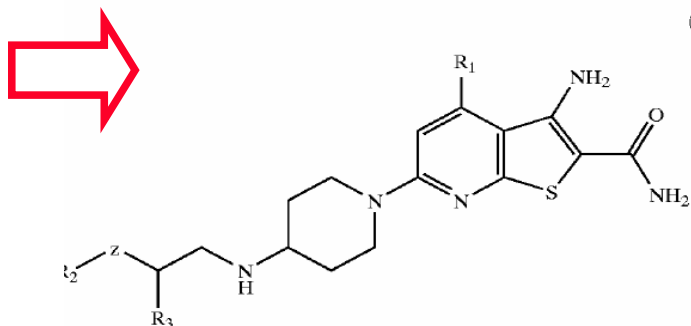
(22) Filed: Dec. 2, 2004

Related U.S. Application Data

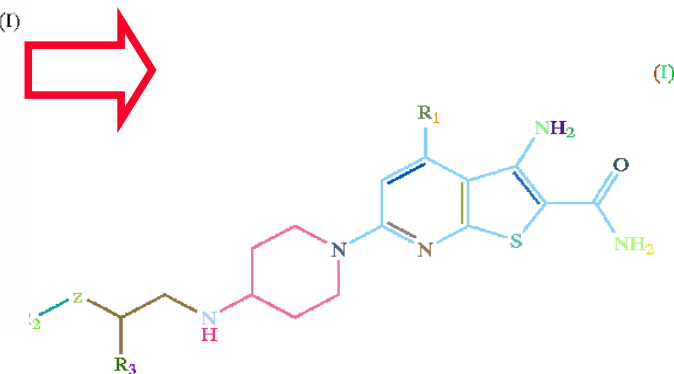
(60) Provisional application No. 60/527,522, filed on Dec. 5, 2003.

wherein the variables R_1 , R_2 , R_3 and Z are described herein, which are useful as inhibitors of the kinase activity of the I κ B kinase (IKK) complex. The compounds are therefore useful in the treatment of IKK mediated diseases including autoimmune diseases inflammatory diseases and cancer. Also disclosed are pharmaceutical compositions comprising these compounds and processes for preparing these compounds.

2 Depiction



3 Reconstruction



4 SDF file

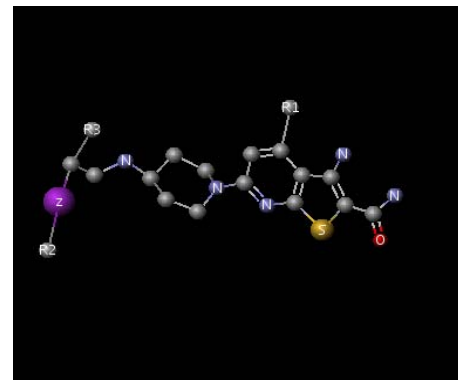
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[illegible]

5 *in silico* Chemistry



Distributed information extraction from scientific literature

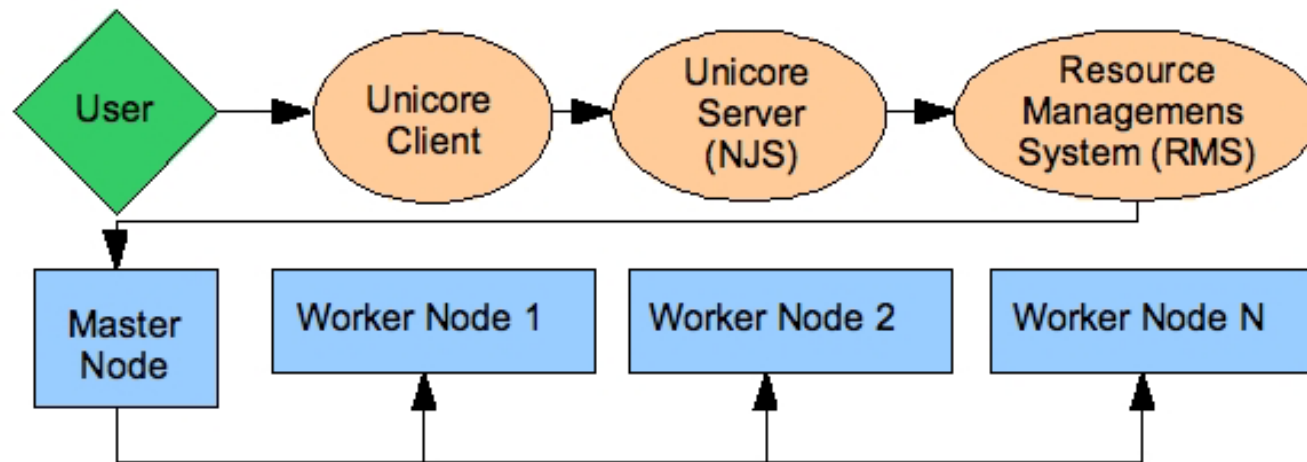
Distributed Documents and Scaling

- ❑ MEDLINE comprises currently more than 16 million abstracts
- ❑ More and more publications available as full text (open access) and in institutional repositories
- ❑ Patent literature comprises more than 50 million full text patents; approximately 13% containing information on chemistry, biology and pharmacology
- ❑ In pharma companies, most relevant information is still available in free text (e.g. information on clinical studies; FDA / BfArM registration)
- ❑ Thousands of news streams and millions of websites comprise valuable information on innovations

Unstructured Information Management Architecture (UIMA)

- ❑ Service Oriented Architecture / framework proposed by IBM
- ❑ Rapidly adopted by commercial and academic tool provider in the area of text mining
- ❑ Supports the assembly of complex annotation workflows
- ❑ Annotators might be entity recognition systems, part-of-speech-analysis modules and other type of unstructured information processing tools
- ❑ UIMA standardizes text and image mining (UIMA not restricted to pure text)

Distributed information extraction from scientific literature



Lessons Learned

- ❑ GRID is an emerging topic for the pharmaceutical industry
- ❑ The WISDOM project has demonstrated, that the GRID is well suited to support large scale virtual screening experiments; making virtual screening a “killer app” for biomedical GRIDs
- ❑ In a completely different field, namely text mining and information extraction, the GRID will enable us to deal with both, distributed documents and compute-intensive tasks
- ❑ Text Mining on the GRID might be the next biomedical GRID “killer app”

Thank you for your attention