

Multiscale simulations of neuronal receptors

Institute for Advanced Simulation

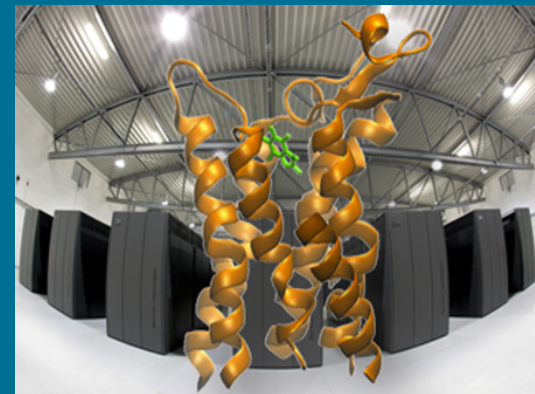
Institute for Neuroscience and Medicine (IAS-5/INM-9/INM-1)

Computational Medicine

Juelich Research Center, Germany

Physics Department, RWTH-Aachen University,
Aachen, Germany

July 3,, 2017| Paolo Carloni



Forschungszentrum Jülich (Jülich Research Centre)

**Interdisciplinary research center,
among the largest in Europe,
focusing on the study and
applications in the areas of health,
neurobiology, information,
environment, and energy**

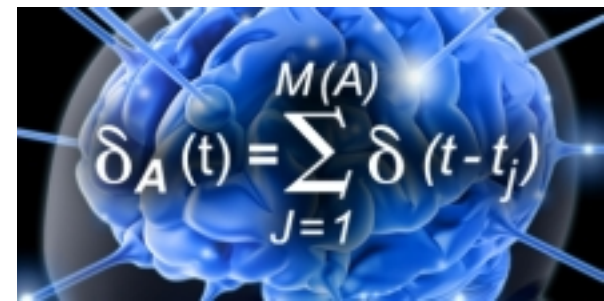
About 1500 scientists (600 PhD's)



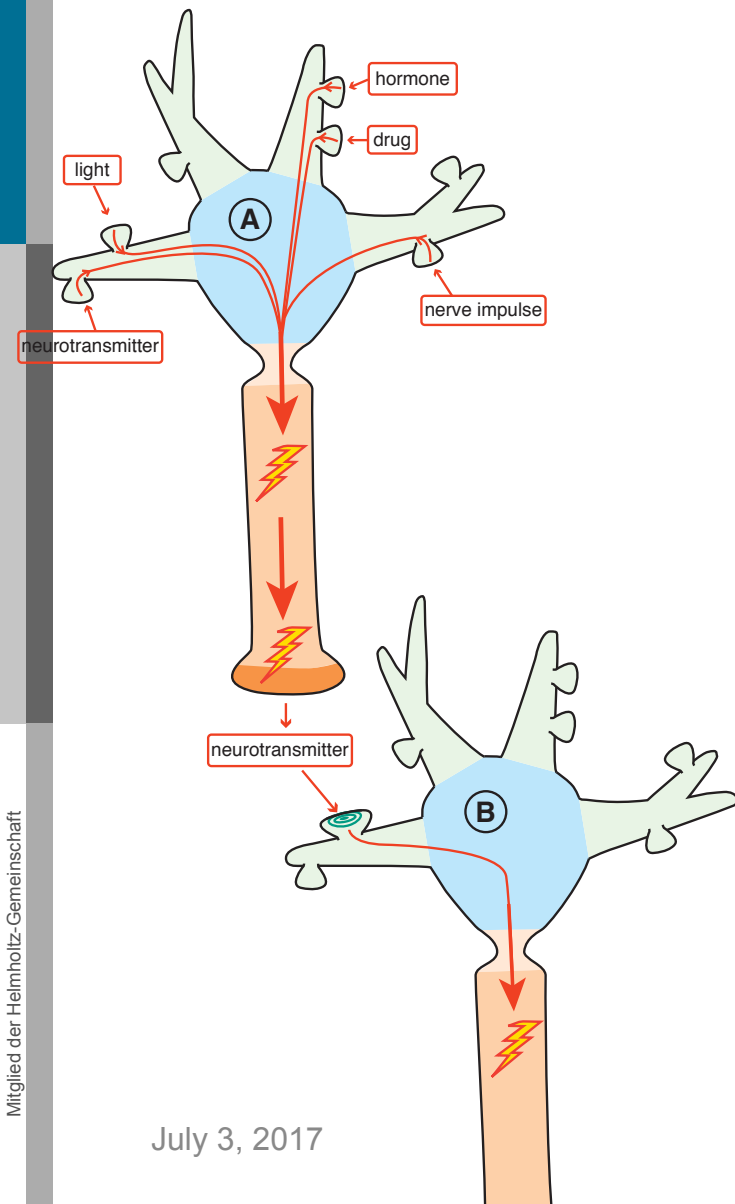
Mitglied der Helmholtz-Gemeinschaft



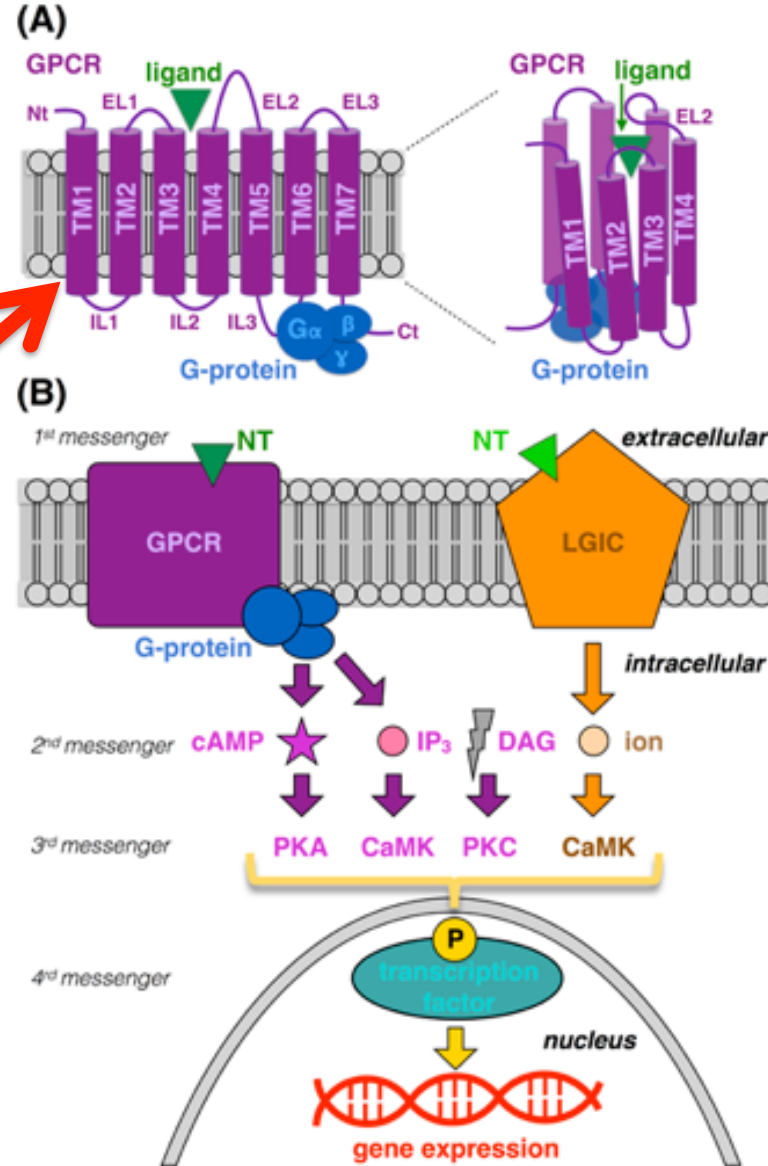
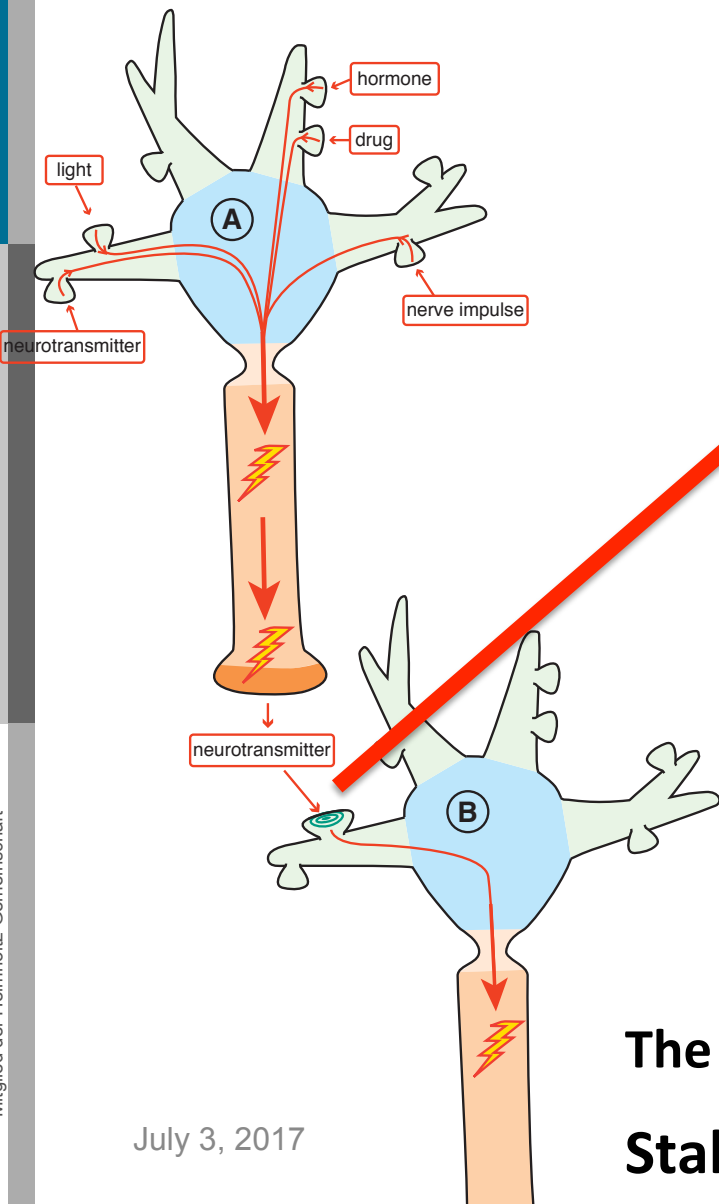
**HELMHOLTZ
| GEMEINSCHAFT**



Synaptic Neurotransmission



Synaptic Neurotransmission: G-protein coupled Receptors



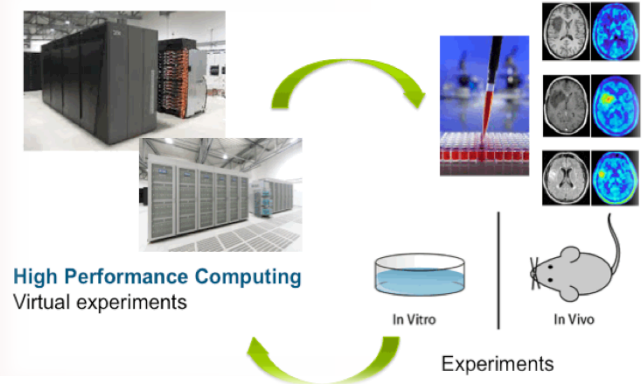
The brain as sophisticated “chemical soup”

Stahl, Essential Psychopharmacology, 2013⁴

Computational biomedicine group

Strategy

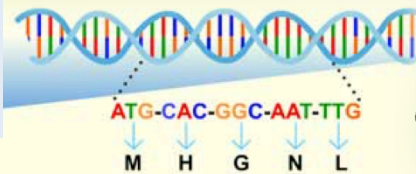
- ➔ Key Molecular events for neuronal function/dysfunction
- ➔ Designing ligands



Biology-based approaches

Tools

From DNA to proteins



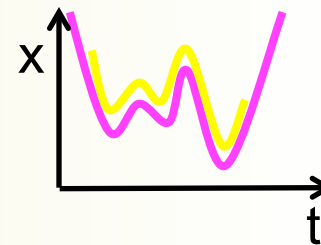
Sequence Compared to Other Organisms

M	H	G	N	LHuman
Q	H	G	N	WMouse
G	H	G	N	KYeast



Physics-based approaches

$$m_i \frac{d^2}{dt^2} \vec{r}_i = -\frac{\partial}{\partial \vec{r}_i} U(\vec{r}_1, \vec{r}_2, \dots, \vec{r}_N)$$



Statistical Mechanics

$$A = -RT \ln Q$$

binding affinities
rate constants

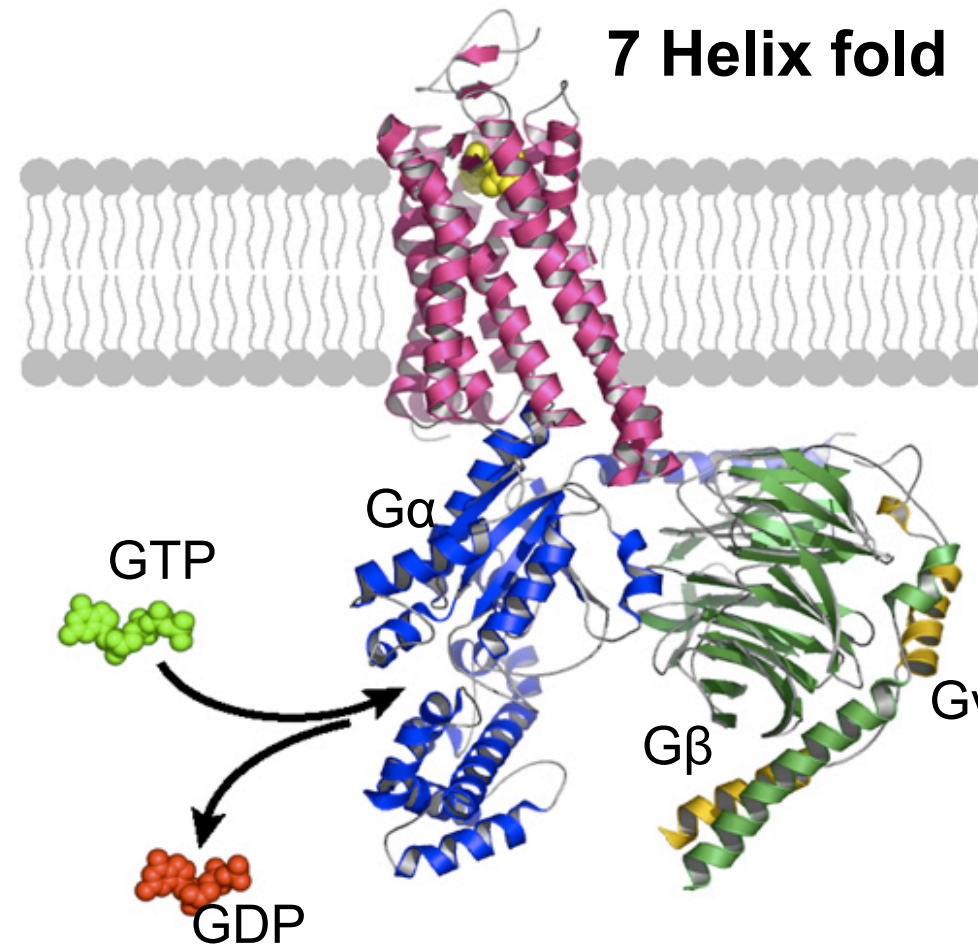
Supercomputing



highly scalable codes for molecular simulation

G-protein Coupled Receptors (GPCRs)

~3%
of mammalian genes
~80%
signal transduction
pathway
across cell membrane
> 30 %
of marketed drugs



Overington et al.
Nat. Rev. Drug. Discov. 2006

Nobel Prize in Chemistry 2012

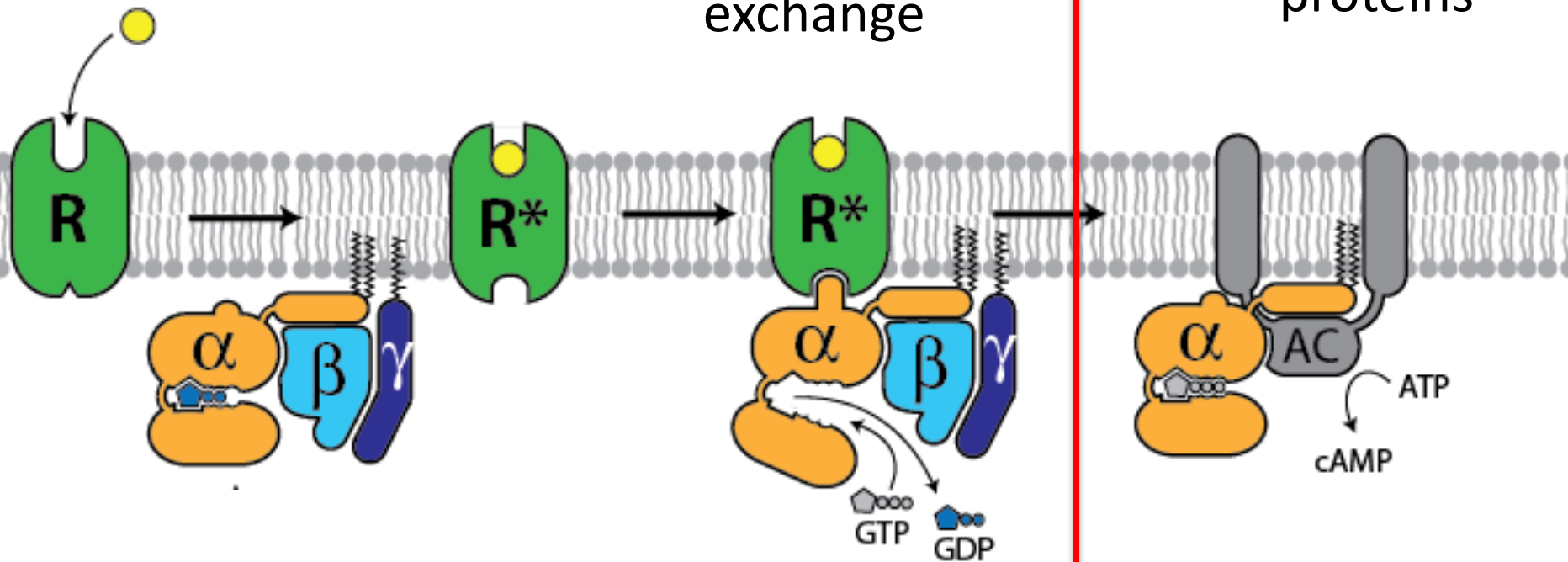
Brian Kobilka (shared with Robert J. Lefkowitz) for structural studies on GPCRs

G-protein Coupled Receptors (GPCRs)

Agonist binding

G protein coupling
and nucleotide
exchange

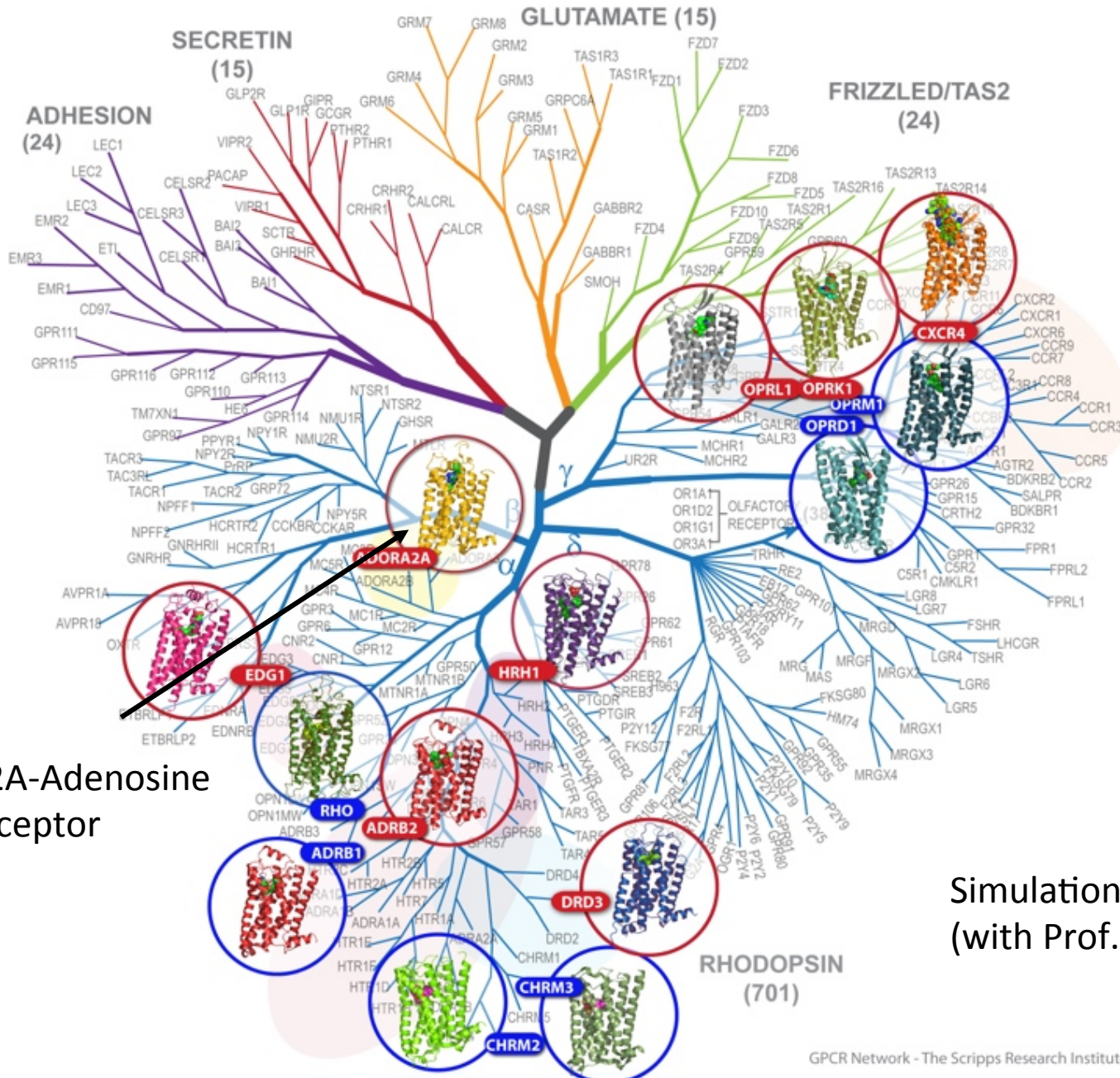
Activated G
protein subunits
regulate effector
proteins



Mitglied der Helmholtz-Gemeinschaft



GPCR structural biology



**~800
members
in the
human
genome.**

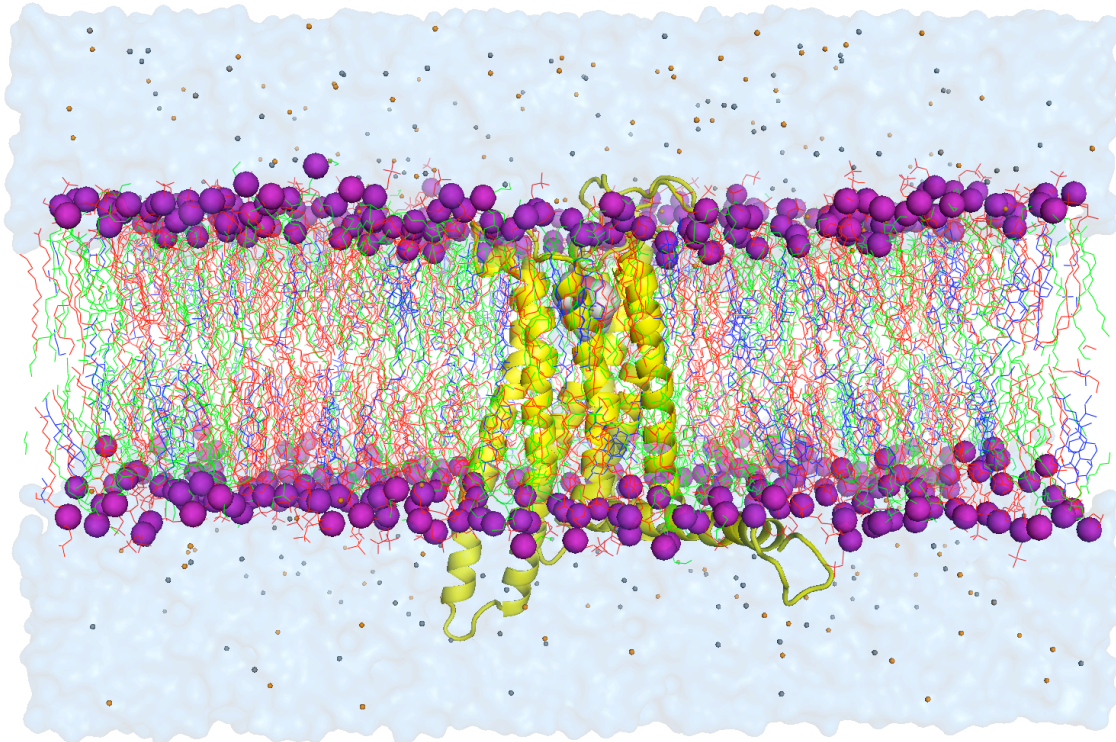
**A2A-Adenosine
receptor**

**Simulation of ligand binding
(with Prof. Bauer, FZJ)**

Molecular dynamics

Simulations on

The hA_{2A}R neuronal receptor



Giulia Rossetti (INM-9/JSC -
W1 Aachen)

Ruiin Cao, with Prof. Bauer
(INM-2)

and Neumeier (INM-5)

Cholesterol 33-50%

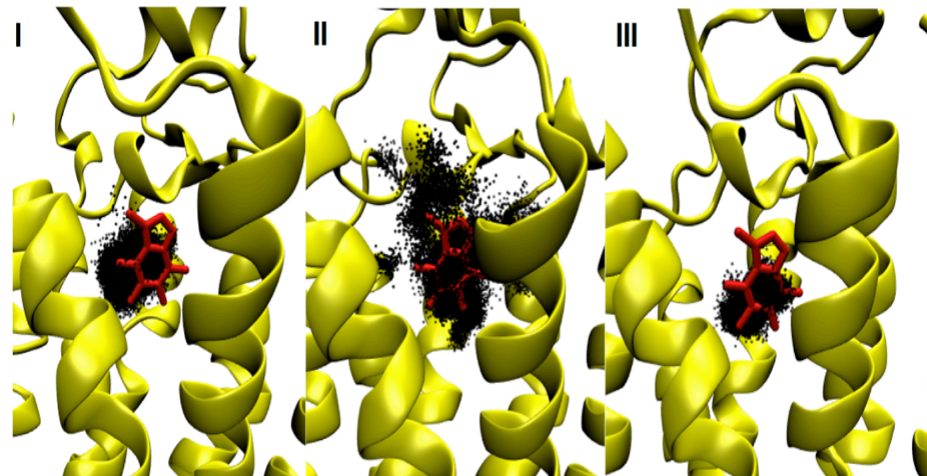
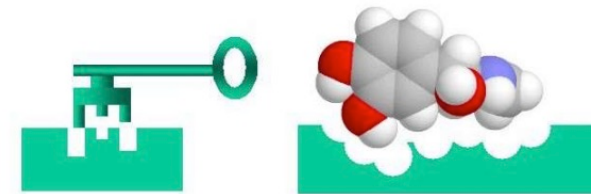
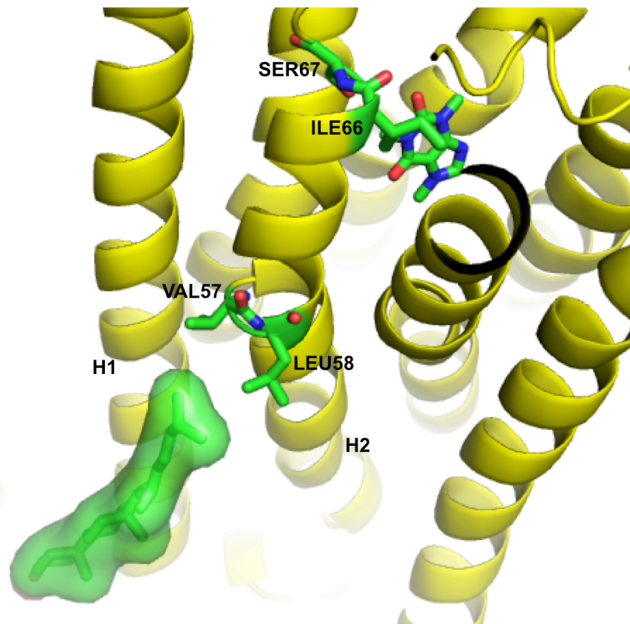
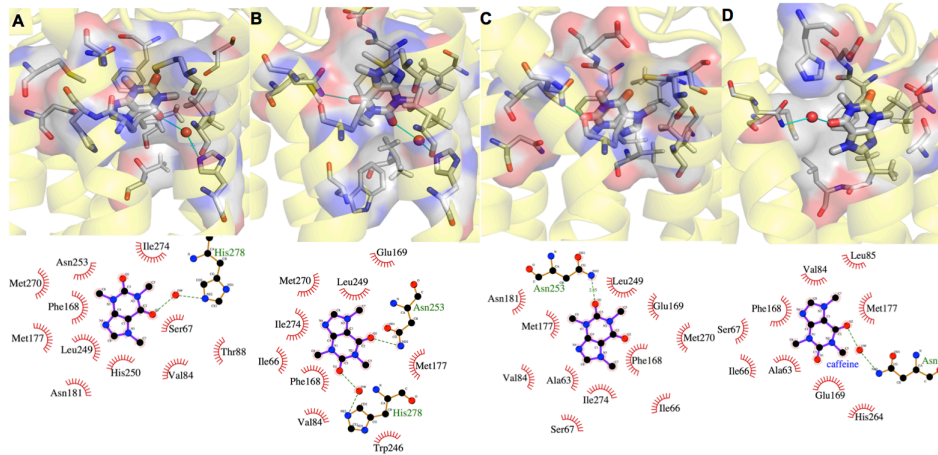
Pfriege, *Biochimica et Biophysica Acta*

PDBid: 3PWH

Amber force field

Microsecond MD

Effect of cholesterol



Ruin Cao et al., *PlosONE* 2015

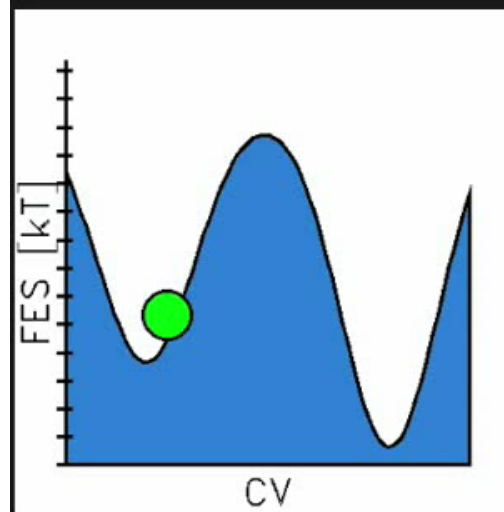
Enhanced sampling for ligand poses and affinities

*A couple of Metadynamics
Reviews from our group*

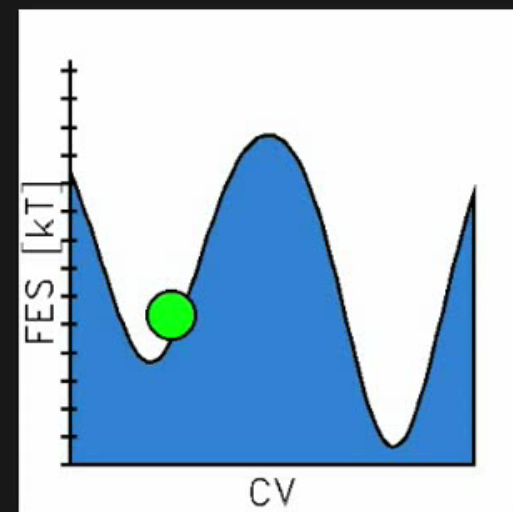
Leone et al. *M. Curr. Opin.
Struct. Biol.*, **2010**

Biarnés et al. *J. Comp.-Aided
Mol. Design*, **2011**,

no metadynamics

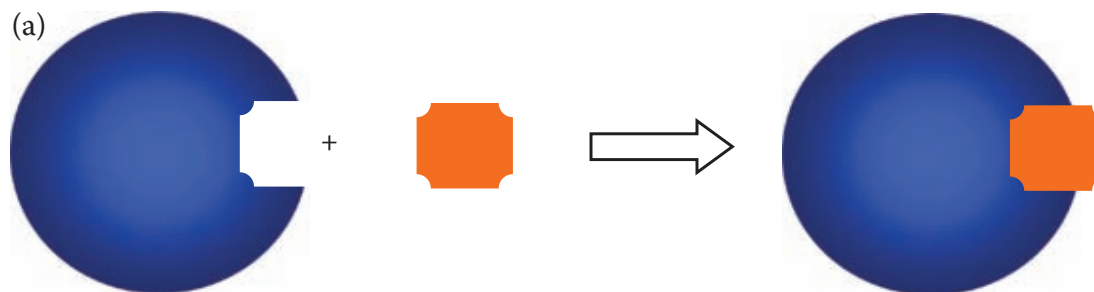


plain metadynamics

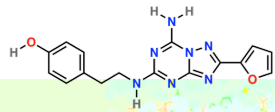


Laio and Parrinello,
PNAS (2002)

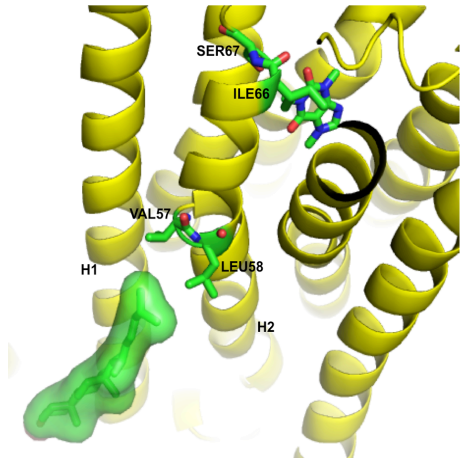
movies by Giovanni Bussi



Ligand poses and affinities of a high affinity ligand of hA_{2A}R



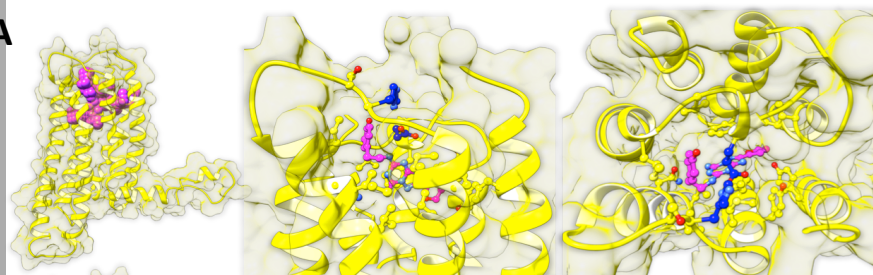
ZMA



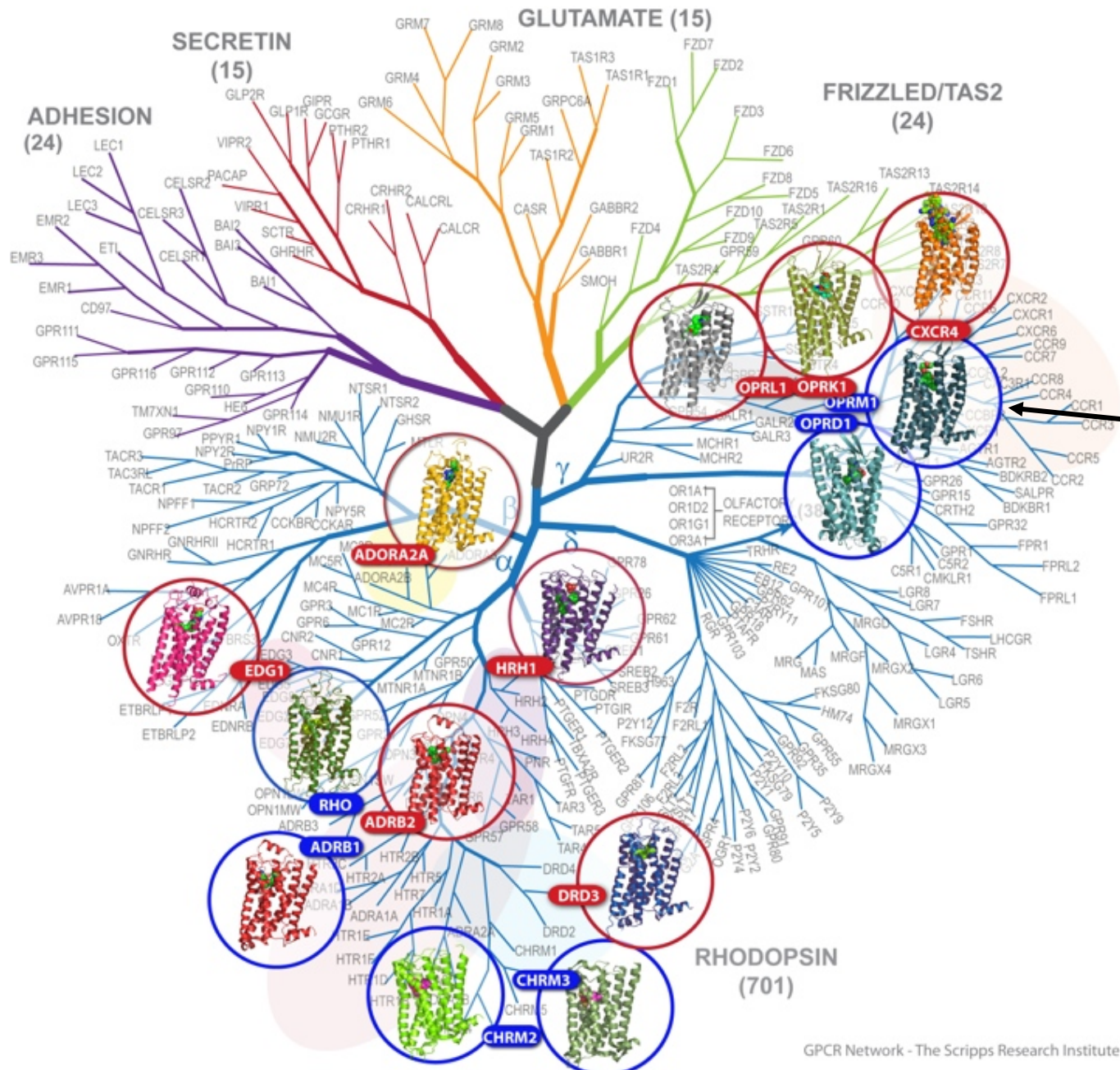
ΔG^0 (exp) = -13.2 kcal/mol
Guo et al. *Mol. Pharm.* 2016

ΔG^0 (calc) \approx -13.9 (8) kcal/mol

Cao et al, *PlosONE* 2015, submitted

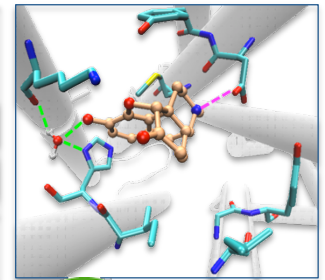
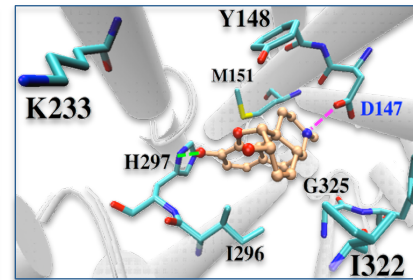
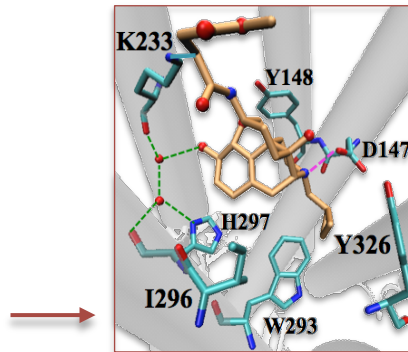
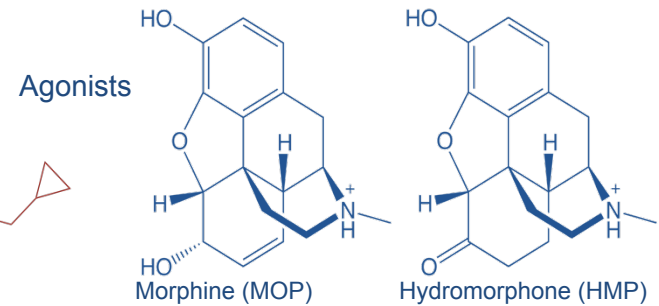
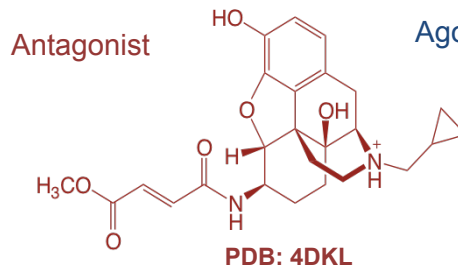
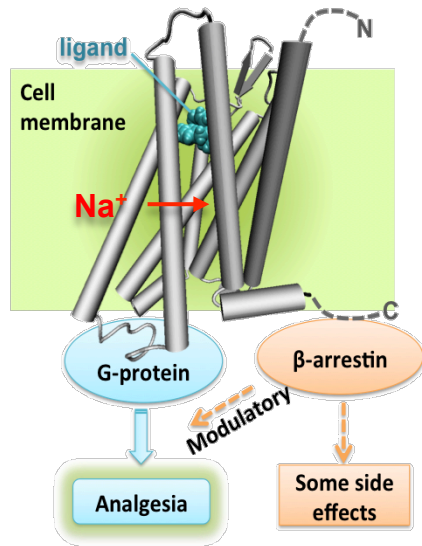


GPCR structural biology

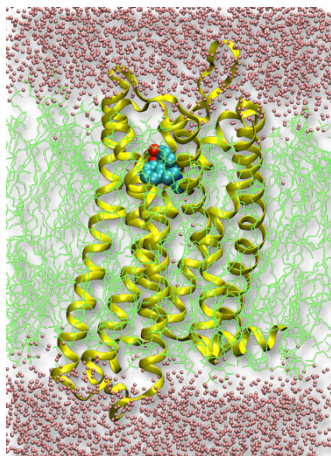


μ-Opioid
receptor

Painkillers targeting the μ -opioid receptor

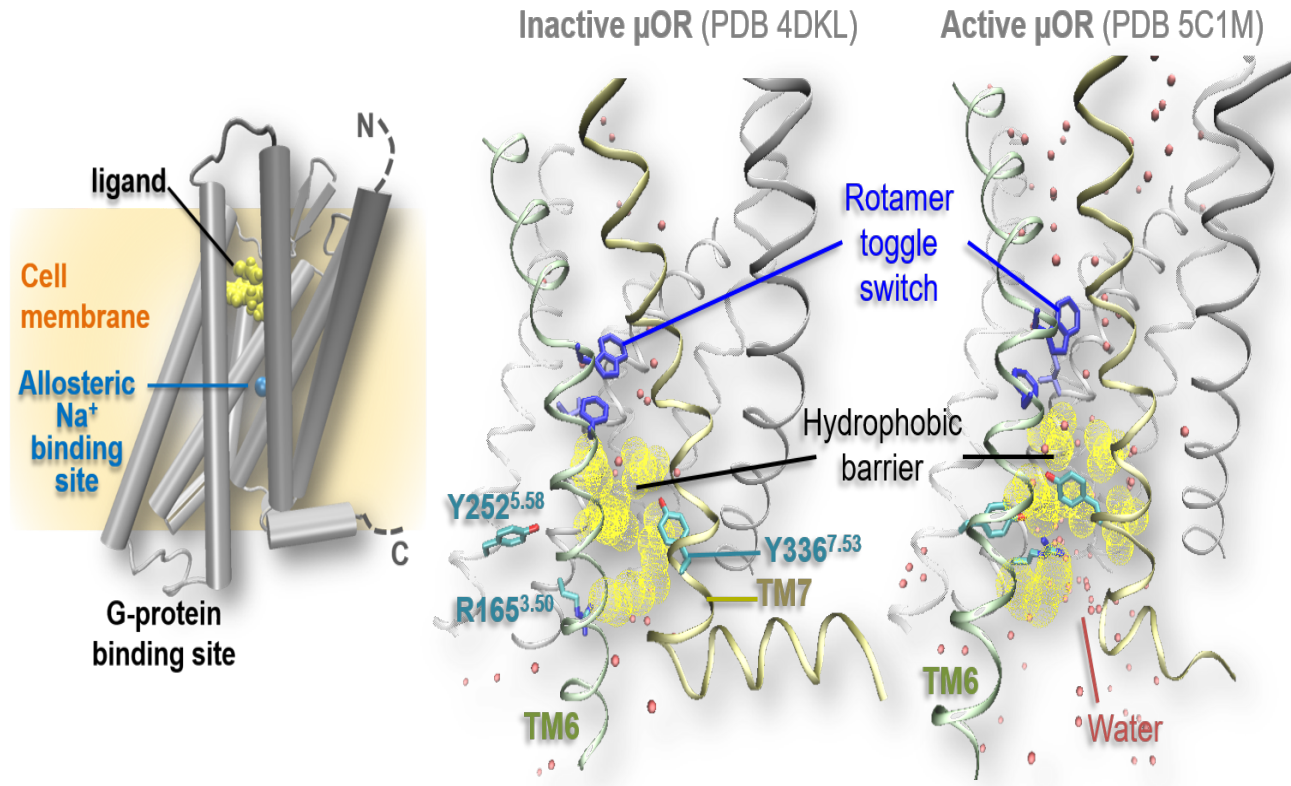


GRÜNENTHAL



**Binding free energy calculation
(Alchemical transformation)**
MOP \rightarrow HMP: 1.2 ± 1.1 kcal/mol
HMP \rightarrow MOP: 0.8 ± 0.8 kcal/mol Exp.
 0.4 ± 0.3 kcal/mol

Activation



force fields:

- AMBER99SB-ILDN - **protein**
- Slipids – **membrane**
- TIP3P - **water** molecules

- PME for electrostatics
- NPT ensemble

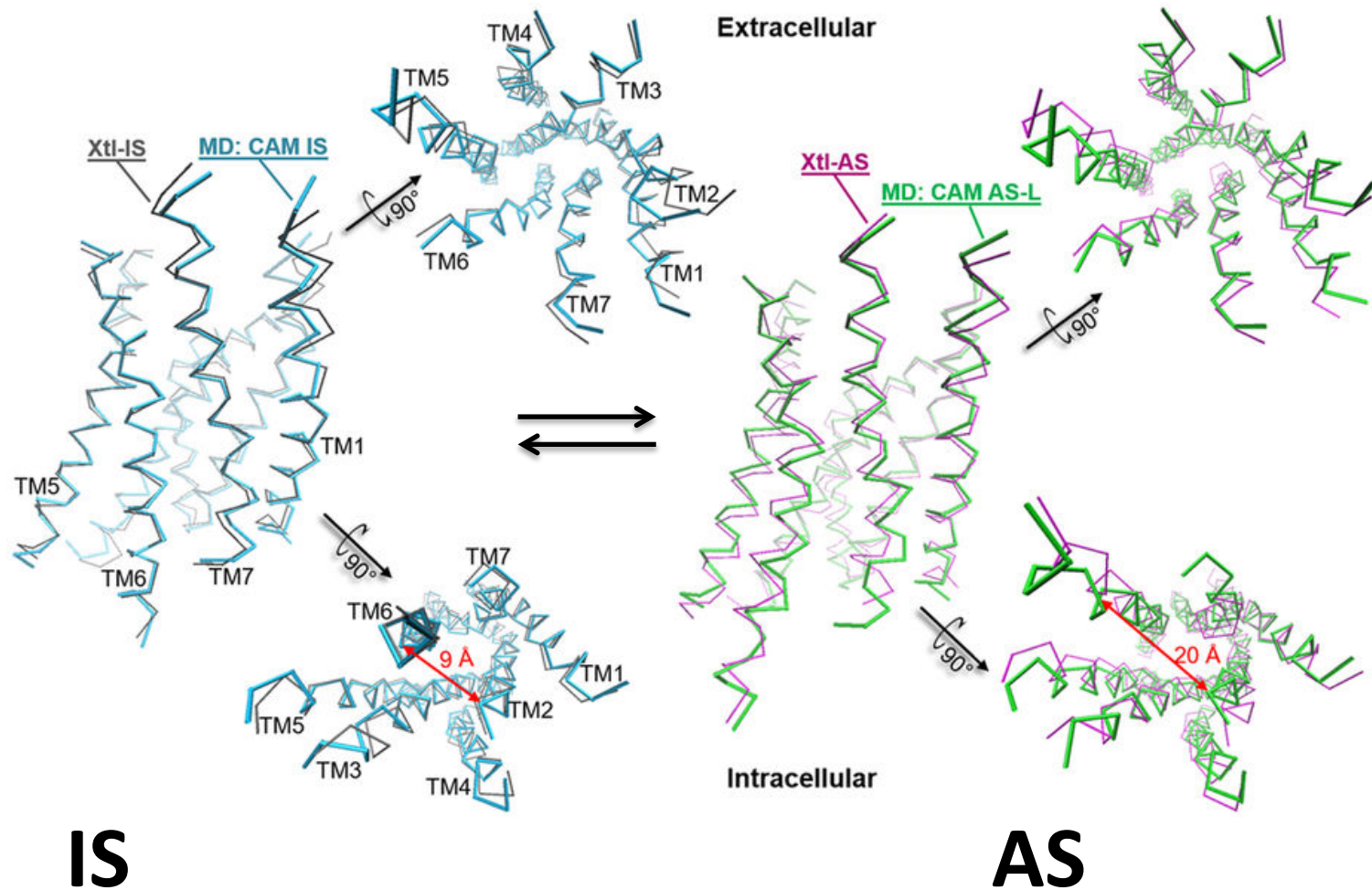
- REST2 : 20 ns \times 64 replicas of MD simulations

- Wang, L., Friesner, R. A. & Berne, B. J. Replica Exchange with Solute Scaling: A More Efficient Version of Replica Exchange with Solute Tempering (REST2). *The Journal of Physical Chemistry B* **115**, 9431-9438.

-biology: *apo* μ OR N150^{3.35} A CAM

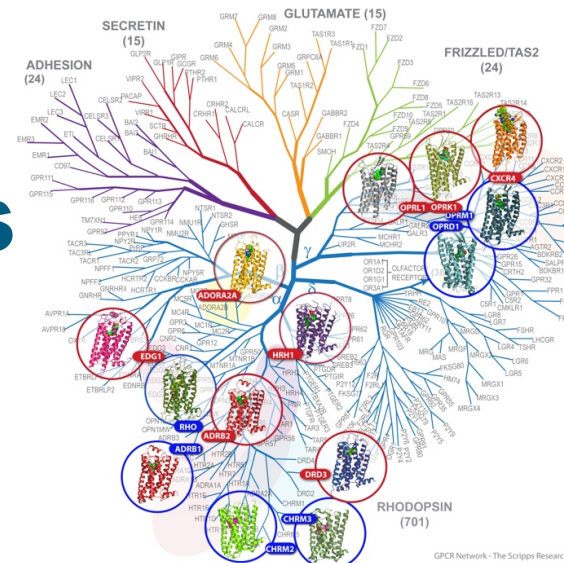
-enhanced sampling: REST2

Interconversion IS (95%)- AS (5%) via few intermediates (not shown)



Cong et al, Sci. Rep., 2017

Largest GPCR's families: Bitter and odorant receptors



TAS2R family
25 members in humans
recognizing hundreds of
food compounds

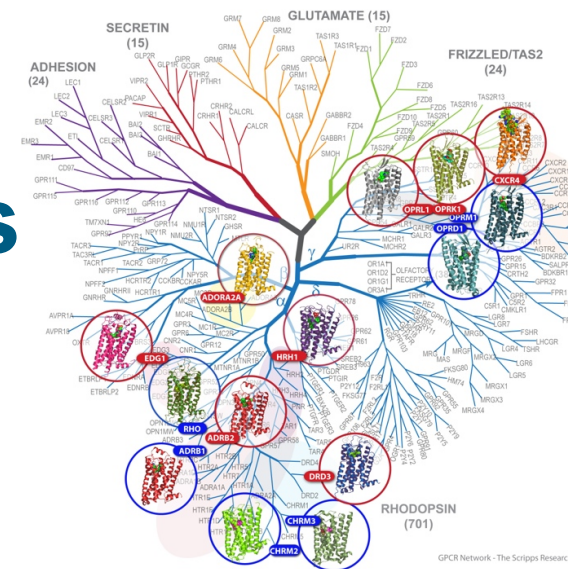
Others (TAS2R38): few agonists binding
Some (TAS2R46): large agonist diversity
access control “for wrong bitter compounds?”

Meyerhof et al. PNAS 2010



Largest GPCR's families: Bitter and odorant receptors

TAS2R family
25 members in humans
recognizing hundreds of
food compounds



Predicting binding poses?

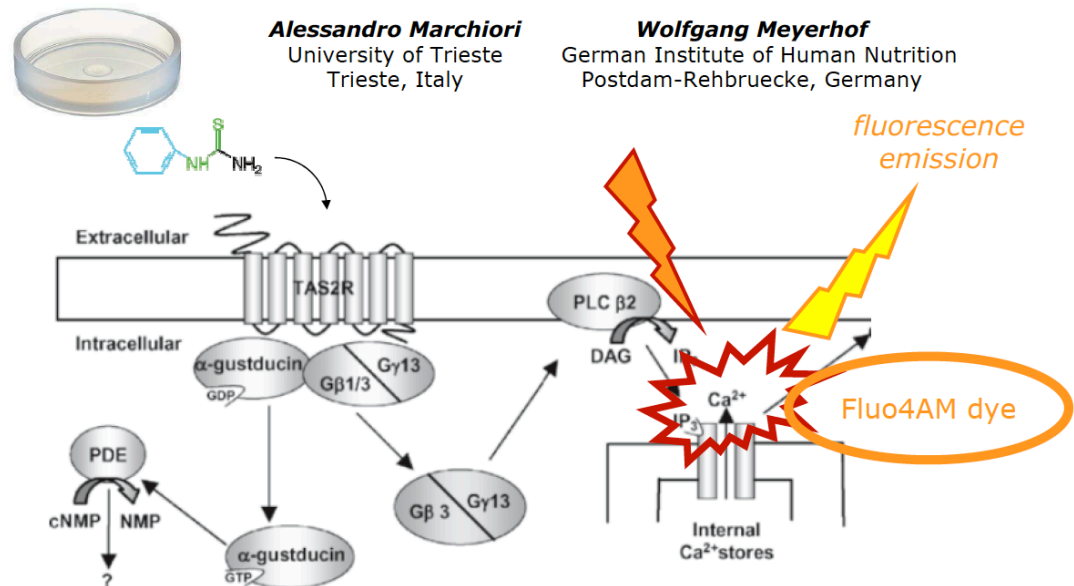
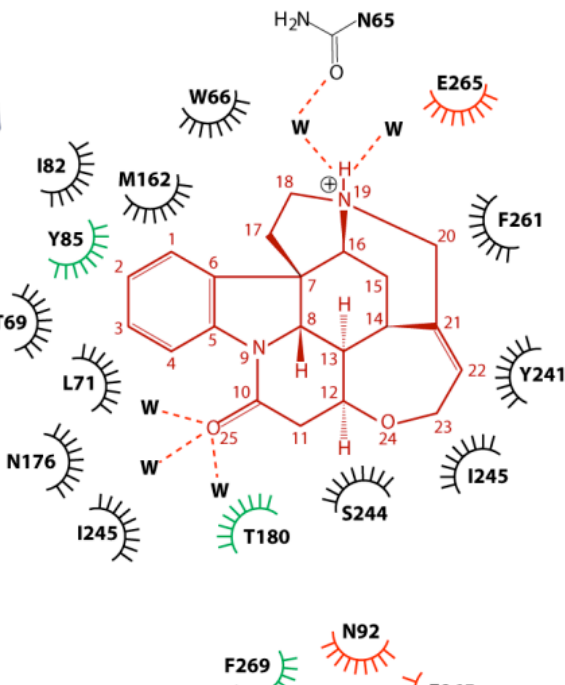
1. No structural information
2. SI with templates <20%
3. Only molecular biology experiments (Meyerhof's group)

Essays on wild-type and mutant bitter taste receptors

TP= # true positives , FP=# false positives
FN= false negatives

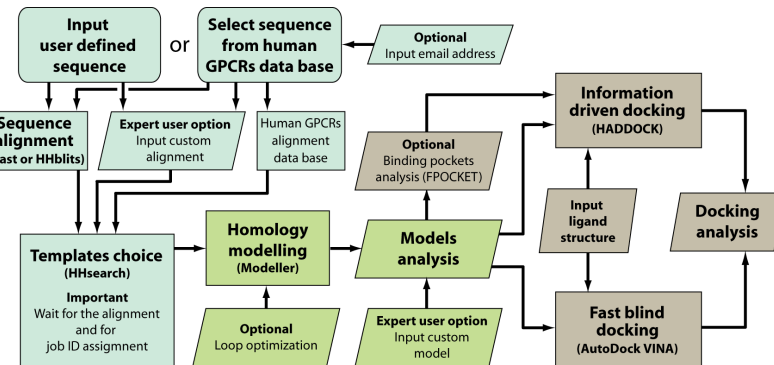
$$\text{PREC} = \text{TP} / (\text{TP} + \text{FP})$$

$$\text{REC} = \text{TP} / (\text{TP} + \text{FN})$$



Step #1: Bioinformatics + docking

1. Available Structural Data
2. Evolutionary Information
3. Sequence based structural/
functional assessment



Structural Modeling of ligand/GPCR complexes

GOMoDo
GPCR Online Modeling and DOcking server

Email address or job code:

Modeling GPCRs
beta version

If this is your first time on GOMoDo, please [read the manual](#) before starting. For an example of modeling results, enter modeling_example in the "Check result" box

Job label:

Your email: You can use it to retrieve your jobs later

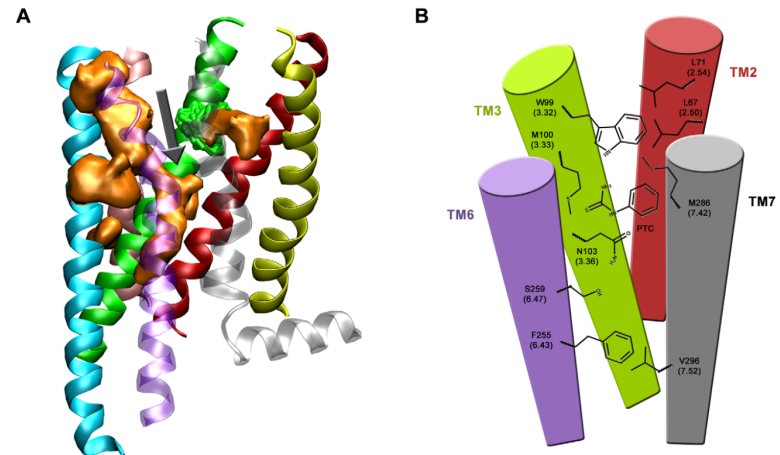
Sequence ☒ Human GPCR

Sequence in FASTA format

Double-check your sequence before submitting !
A define (the line beginning with ">") is required.

```

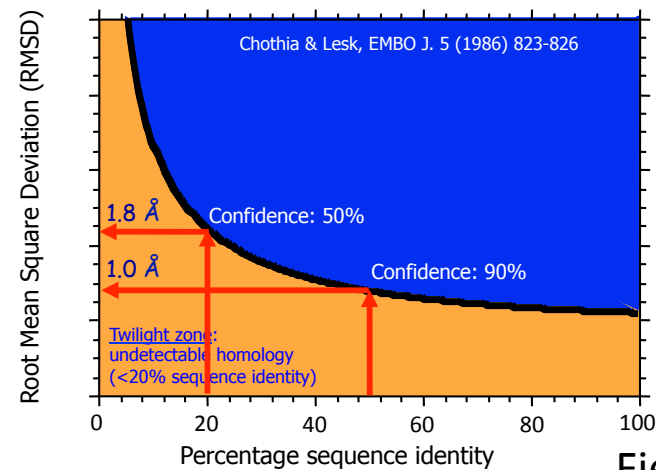
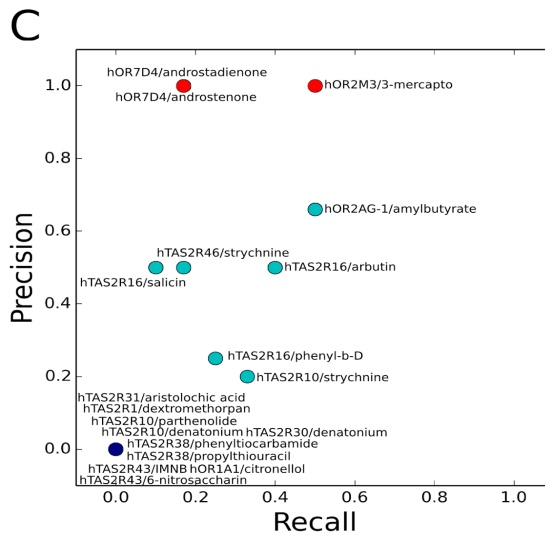
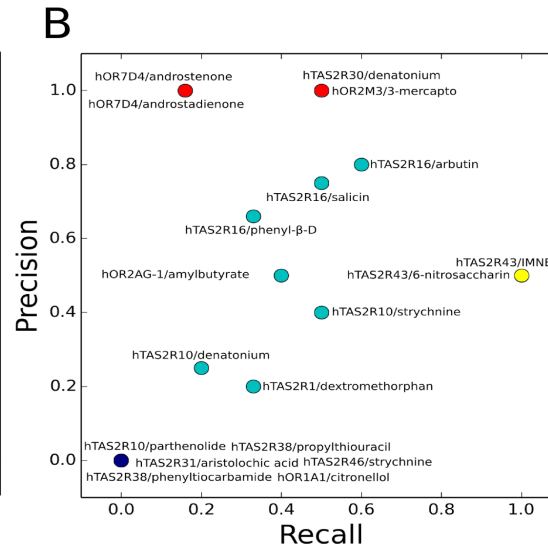
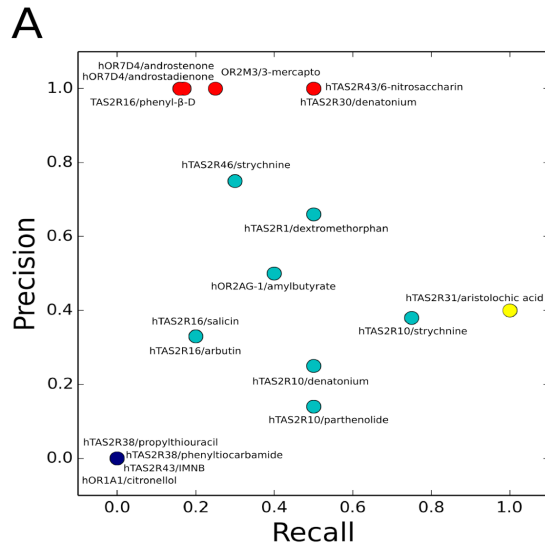
>sp|P59533|T2R38_HUMAN Taste receptor type 2 member 38 OS=Homo sapiens GN=TAS2R38 PE=2 SV=3
MLILIRIRIVVEVRSITFLFISVLEFVAGFLINAFVFNVDVYRQALNSDQVLCL
SISRFISGLFLSAICLTHFGSEFLHWSVQAIQCHRIANQANWGAACLSLYGH
LIRFSFTFLICLASHVSRFISQMLIILICSCICITVLQWCFSSRPHFTVITLPGNHT
RLNWQIKDNLVYSFLFCYLVSWVPPFLFLVSSQMLTVSLGRDRINKVYTRNSRPSLE
ARIKALISLVFFCFVVISCAAFISVFLILMRDWIGVYGVINACFSGRHAILIS
NMFLEKAVNTILNAGSSRLVADHFAQSRILC
    
```



Alejandro Giorgetti

Kamil Khafizof

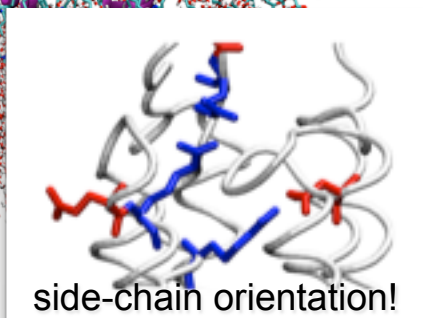
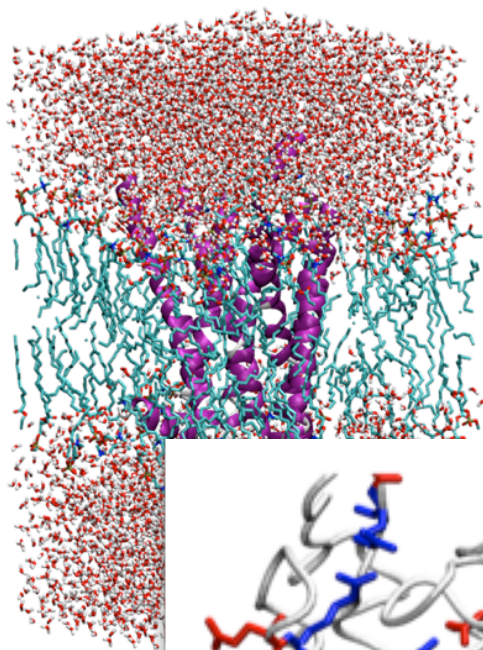
Step #1: Bioinformatics + docking



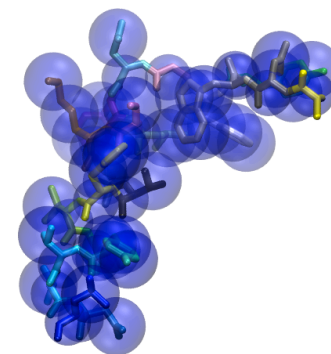
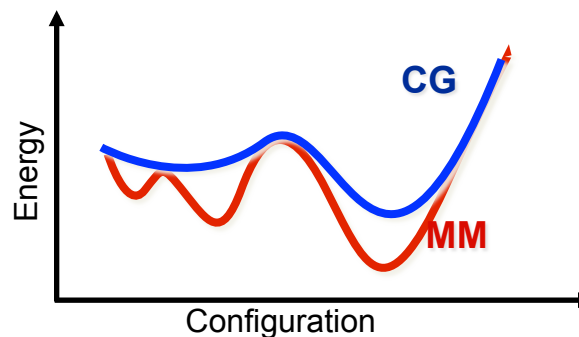
Fierro et al, submitted.

Step #2: Molecular Simulation

all-atom model



coarse grain model



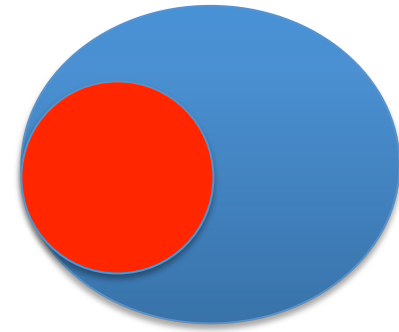
MM: all-atom
CG: coarse-grained

Ligand binding

Hybrid methods

Investigating structure, dynamics and energetics of proteins by molecular dynamics at different levels of granularity (quantum-mechanical, all-atoms force field, coarse grain)

→ Combining different descriptions



Coarse-grain/classical mechanics

Hybrid Coarse-Grain/Molecular Mechanics simulations

PRL 95, 218102 (2005)

PHYSICAL REVIEW LETTERS

week ending
18 NOVEMBER 2005

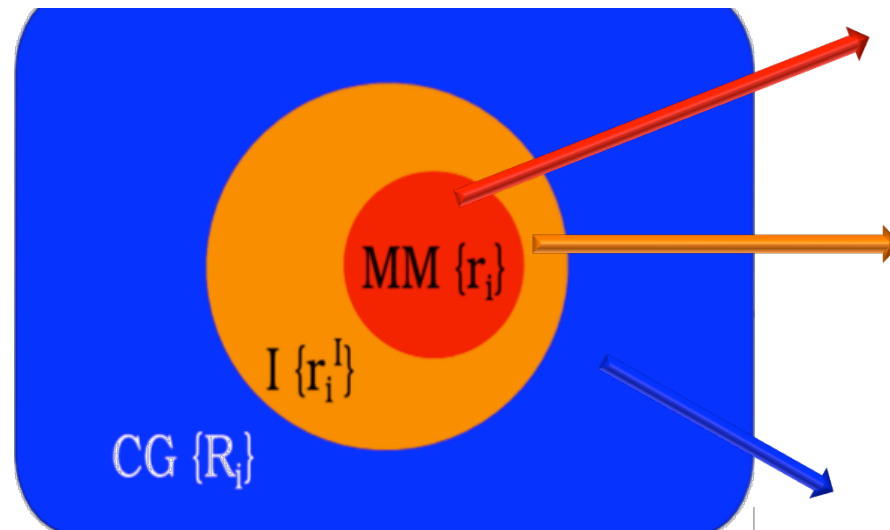
Coarse-Grained Model of Proteins Incorporating Atomistic Detail of the Active Site

Marilisa Neri,¹ Claudio Anselmi,¹ Michele Cascella,² Amos Maritan,³ and Paolo Carloni^{1,*}

¹SISSA/ISAS and INFN-DEMOCRITOS Modeling Center, Via Beirut 4, I-34014 Trieste, Italy

²Ecole Polytechnique Fédérale de Lausanne (EPFL), CH-1015 Lausanne, Switzerland

³Dipartimento di Fisica and INFN, Università degli Studi di Padova, Via Marzolo 8, I-35131 Padova, Italy
(Received 20 April 2005; published 16 November 2005)



MM: Region of interest
(e.g. protein active site)

Atomistic force field (Gromos, Amber)

Interface region

Atomistic force field coupled to CG

CG: Protein frame
Go-model

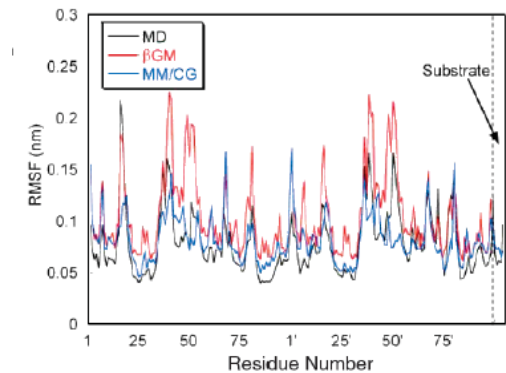
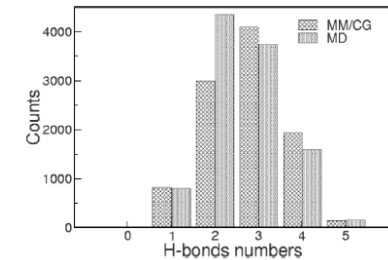
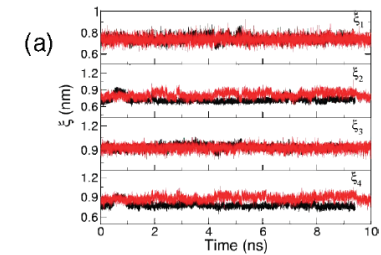
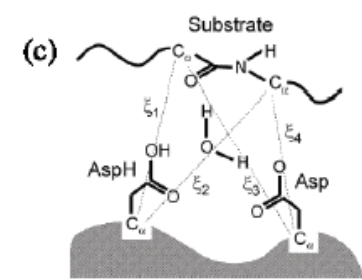
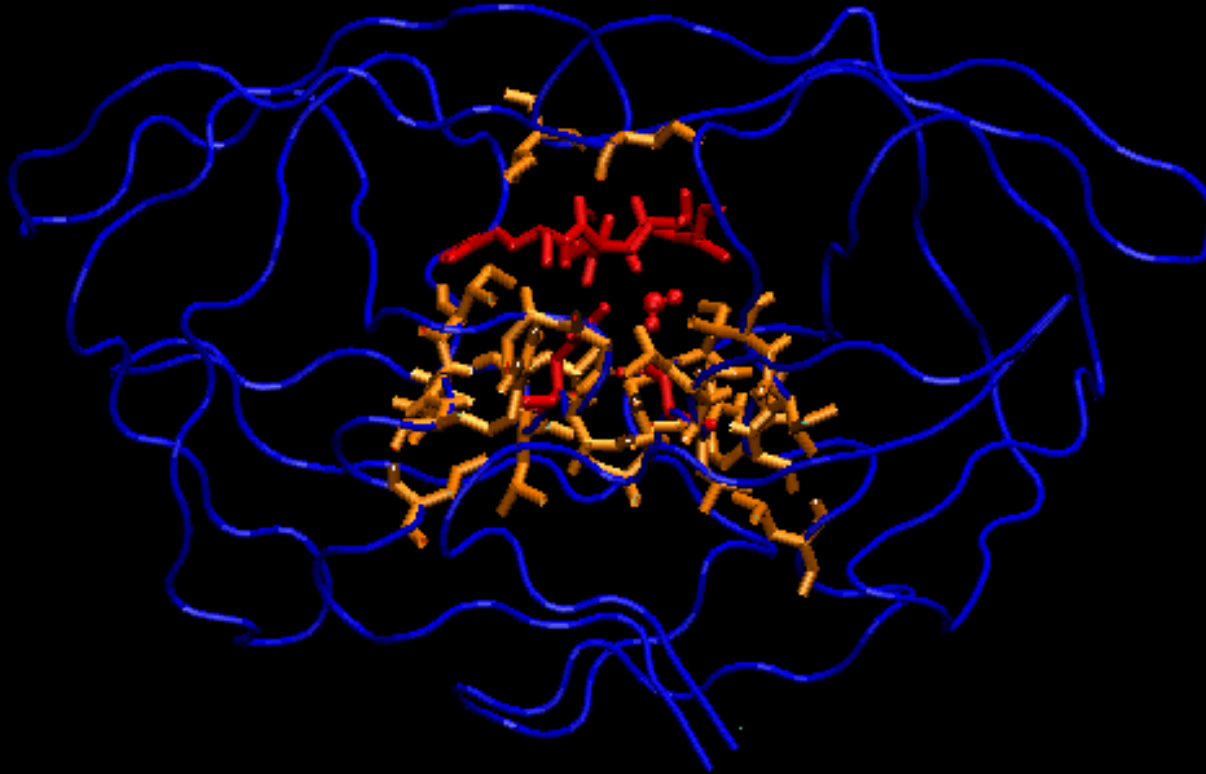
(e.g. Cα backbone atoms only)

$$V = E_{MM} + E_I + E_{I/MM} + E_{CG} + E_{CG/I}$$

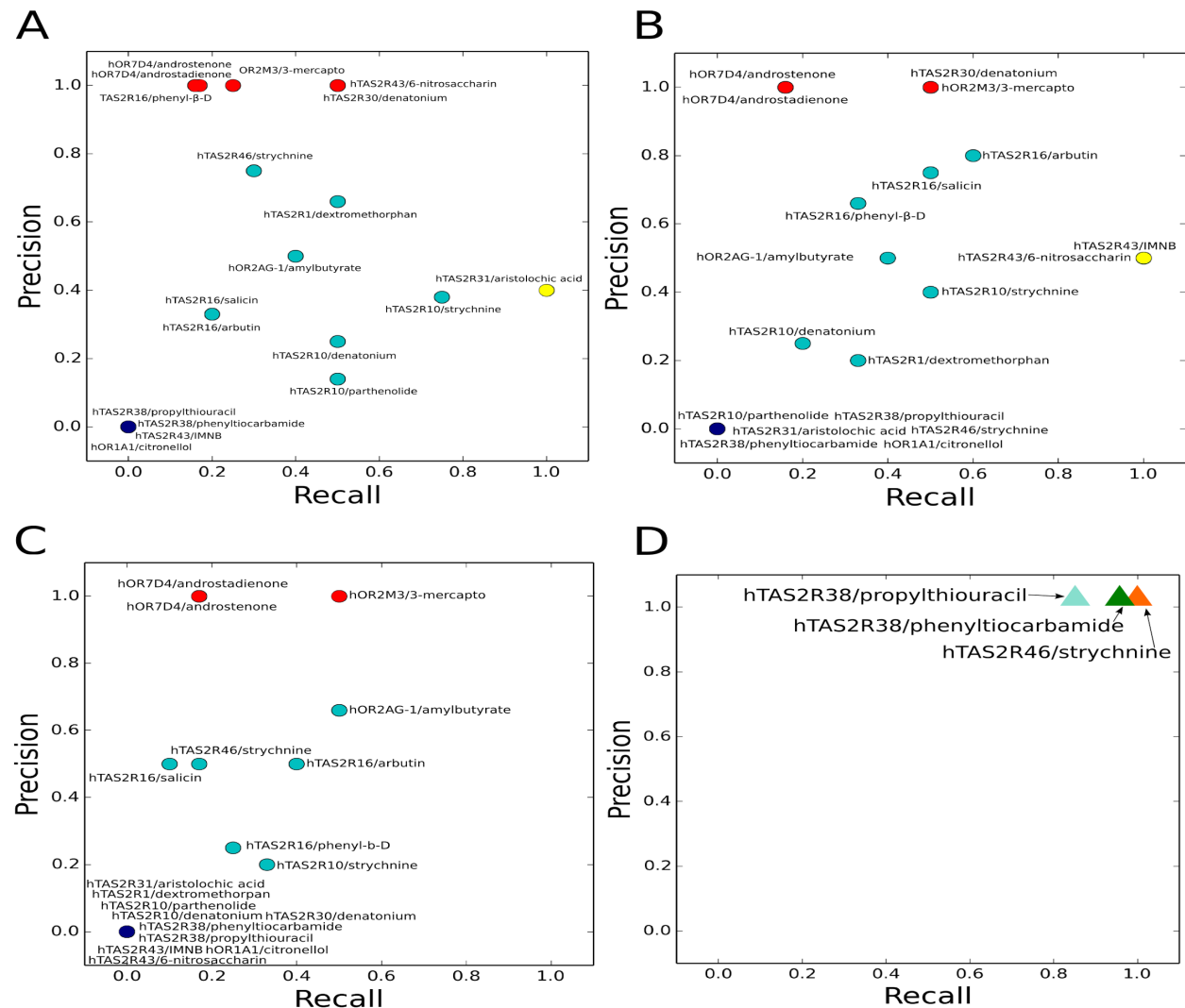
$$E_{MM} = E_{\text{bond}} + E_{\text{vdW}} + \sum_{i>j} \frac{q_i q_j}{\epsilon |\mathbf{r}_i - \mathbf{r}_j|},$$

$$E_{CG} = \frac{1}{4} \sum_i K_b (|R_i - R_{i+1}|^2 - b_{ii+1}^2)^2 + \sum_{i>j} V_0 \{1 - \exp[-B_{ij}(|R_i - R_j| - b_{ij})]\}^2$$

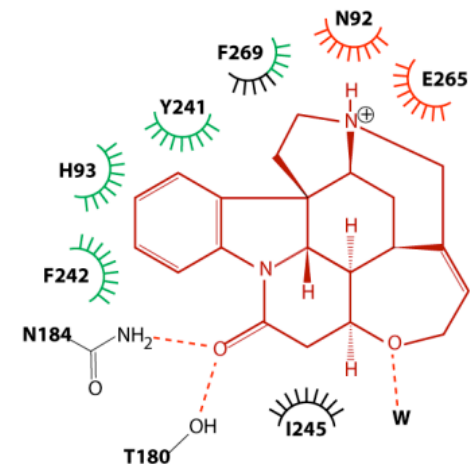
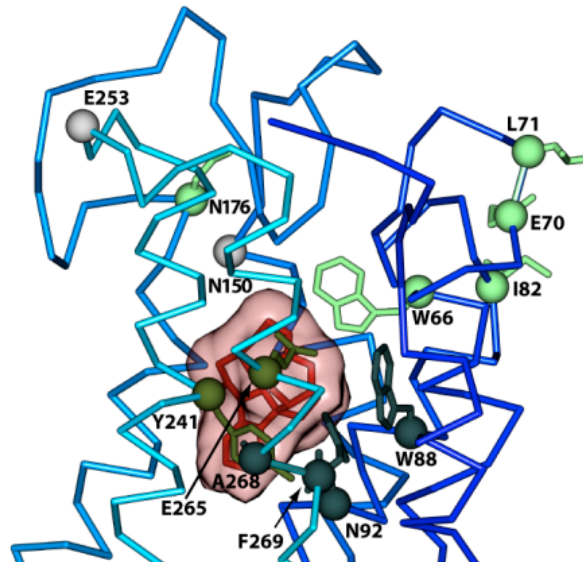
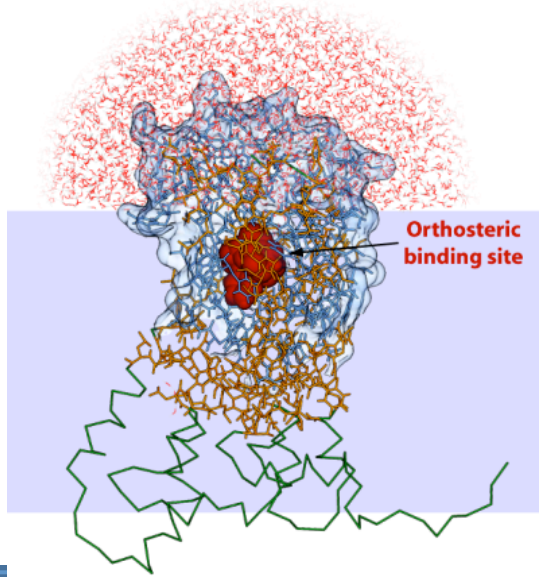
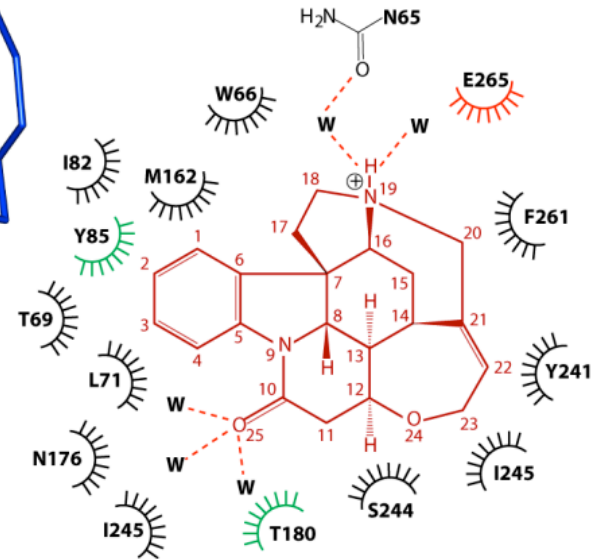
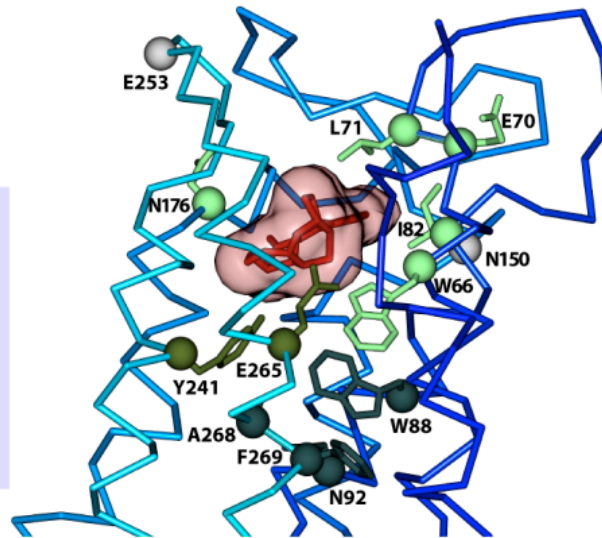
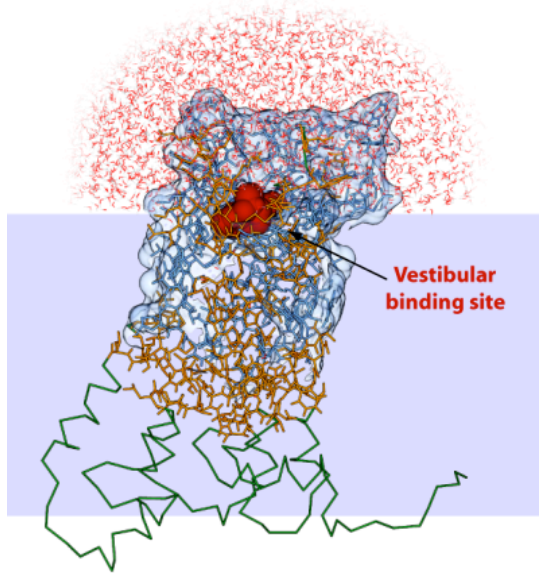
Test #1: A cytoplasmatic protein



Coarse grain/MD of TAS2R46 And TASR38 in complex with agonists



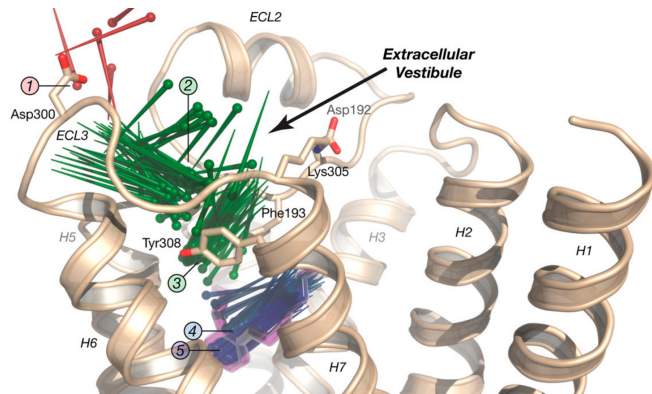
Coarse grain/MD of TAS2R46/ Strychnine (Large ligand diversity)



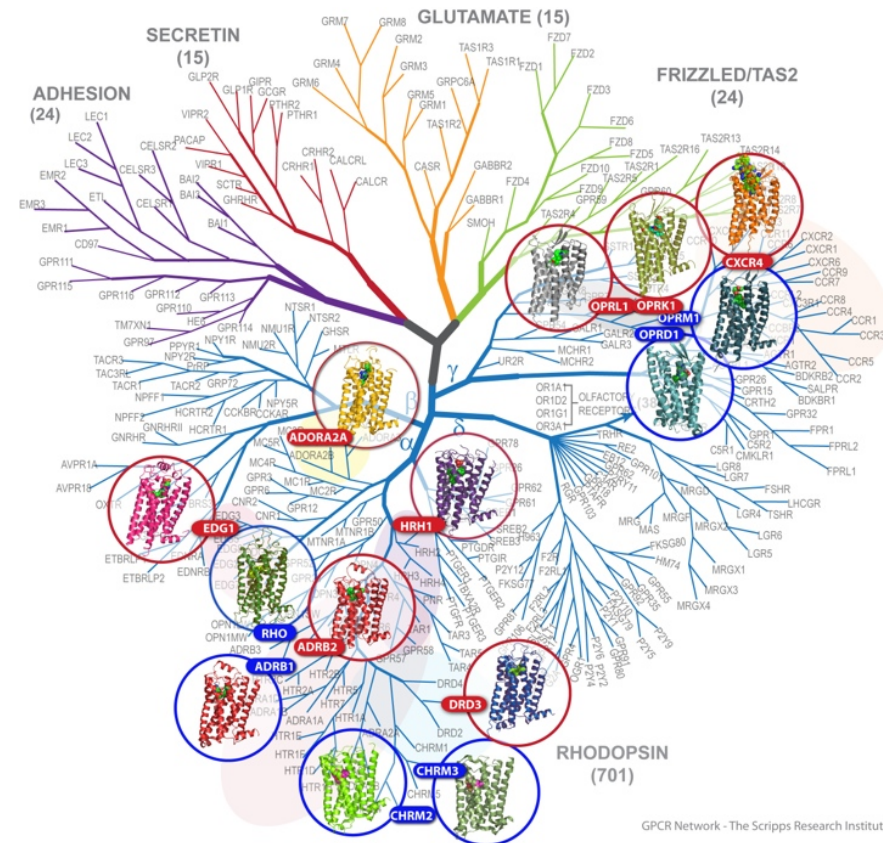
'Vestibular' binding site across

GPCRs

Muscarinic, β 2-adrenergic, opioid receptors: two cavities



alprenolol/ β 2-adrenergic

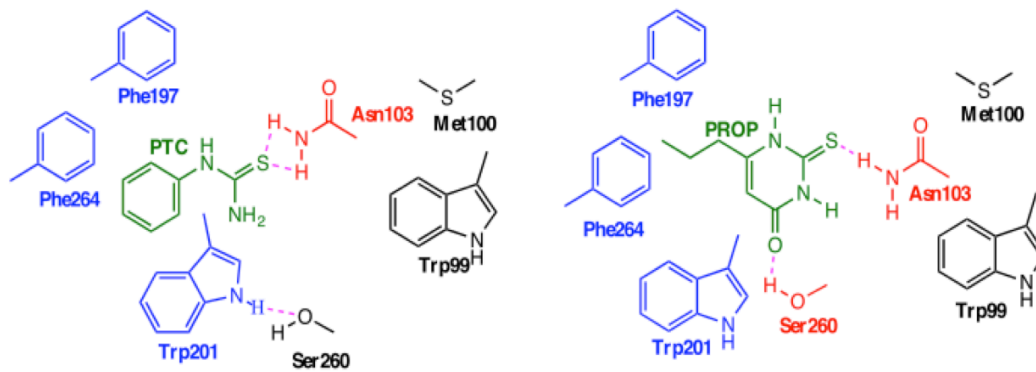


GPCR Network - The Scripps Research Institute

Dror et al. *PNAS* 2011, *Nature* 2013
Granier. & Kobilka, *Nat. Chem. Biol.* 2012

TAS2R8: Only one binding site

- Predictions consistent with more than 20 site-directed mutagenesis and functional calcium imaging experiments



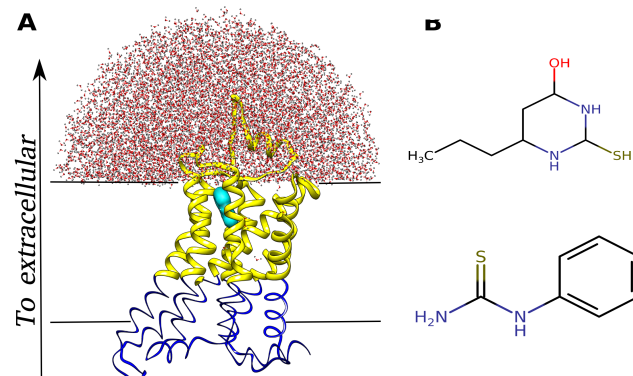
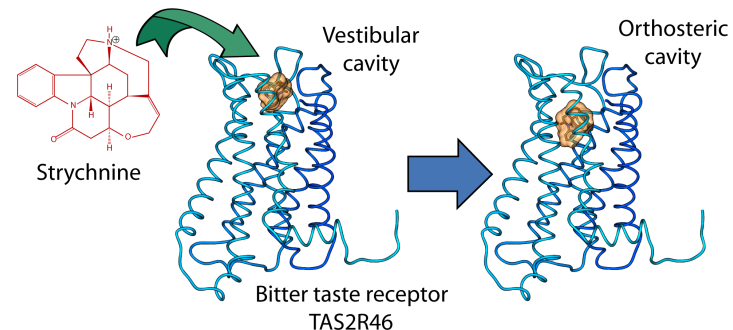
Variant	Agonist			
	PTC		PROP	
	EC ₅₀ (uM)	Max act	EC ₅₀ (uM)	Max act
WT	2.5 (3)	0.43 (0.47)	2.17	0.44
Trp99Ala	1.2 (4.25)	0.14 (0.25)	1.8	0.59
Trp99Val	1.8 (2.7)	0.28 (1.12)	5~	0.93
Met100Ala	4.1 (3)	0.72 (1.01)	1.2	0.77
Met100Val	21.2* (10)	0.51 (0.79)	1.8	0.42
Asn103Ala	6.6* (8)	0.21 (0.38)	8.7*	0.65
Asn103Val	6.9* (15)	0.09 (0.09)	9.1*	0.41
Asn103Asp	-	0.06	23.8*	0.13
Asn179Ala	4.4	0.34	4.9	0.32
Asn179Val	4.9	0.29	5	0.30
Arg181Ala	2.2	0.26	4.3	0.26
Arg181Val	4.5	0.17	7.5	0.19
Asn183Ala	4.2	0.36	5.3	0.32
Asn183Val	2.5	0.44	3.1	0.40
Phe197Val	4.3	0.06	9.9*	0.12
Trp201Leu	-	0.25	-	0.02
Trp201Phe	21*	0.14	-	0.05
Ser259Ala	5.7 (5.4)	0.55 (0.42)	2.9	0.45
Ser259Val	99* (27)	0.02 (0.04)	21.8*	0.18
Ser260Ala	1.41	0.21	1.14	0.45
Ser260Val	9.8*	0.03	6.8*	0.09
Phe264Ala	-	0.06	-	0.06
Phe264Val	12.4*	0.06	25.9*	0.24

Marchiori et al. *PLOS ONE* 2013

Modulating selectivity: from 2-state to 1-state binding

**Input: sequence/
chemical structure
compound**

Agonist structural diversity



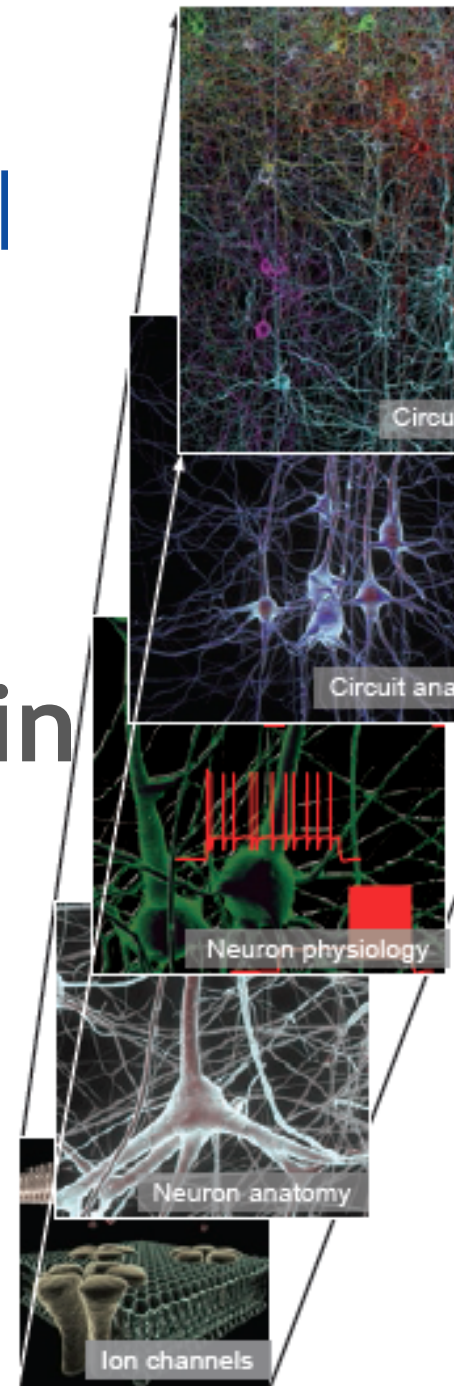
Sandal et al,
J. Chem. Theory Comp. 2015



Human Brain Project



Multi-scale molecular simulation in the human brain project



GPCRs-based signaling pathway involved in memory processes

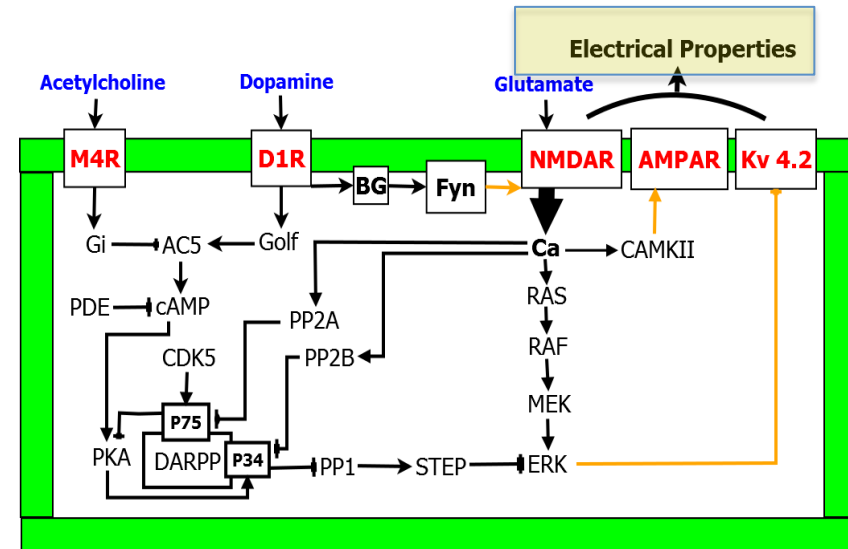
- Case study: Receptor induced cascades leading to activation of kinases and phosphatases, integrated into a neuron model

Proof of concept:

<https://collab.humanbrainproject.eu/#/collab/489/nav/5364>.

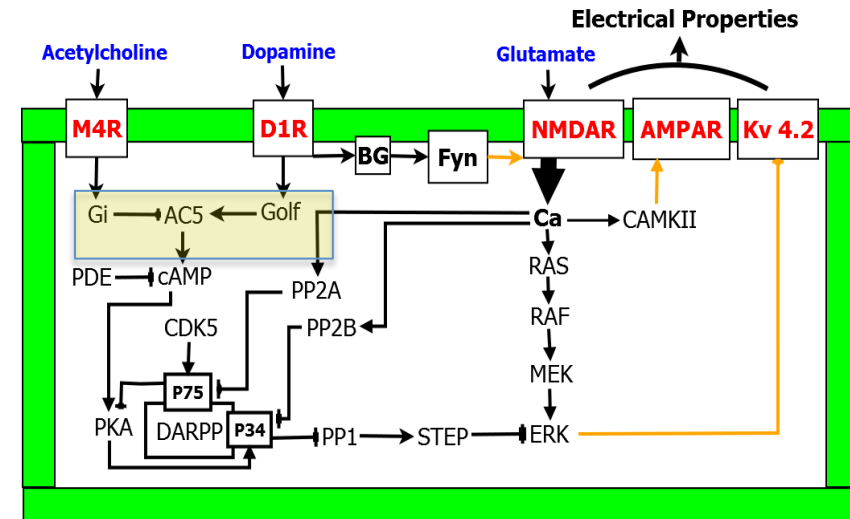
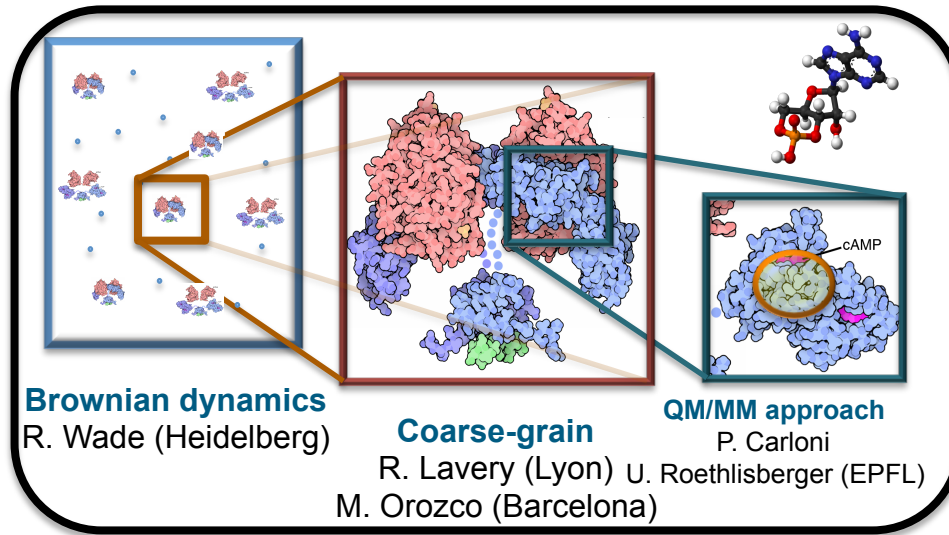
Molecular Dynamics

- Input data for systems biology (at times not accessible from experiment)
- Molecular details of neuronal cascades relevant for higher level models



Nair, et al (2015) J Neuroscience

Multi-scale molecular simulation and systems biology



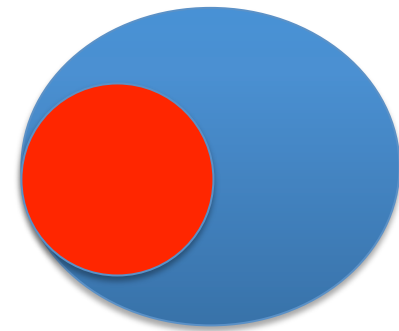
General approach for

- kinetic parameters
- molecular insights

Hybrid methods

Investigating structure, dynamics and energetics of proteins by molecular dynamics at different levels of granularity (quantum-mechanical, all-atoms force field, coarse grain)

→ Combining different descriptions



Quantum mechanics/classical mechanics

QM/MM simulations

CPMD code

DFT

BLYP, BP, B3LYP XC functionals

Basis set: PW (90 Ry)

Martins Troullier pseudopotentials

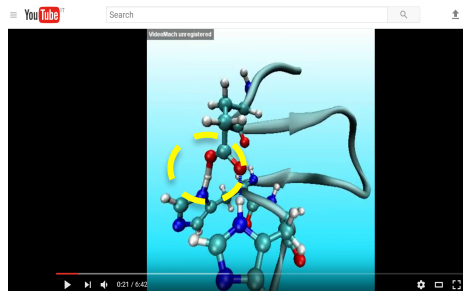
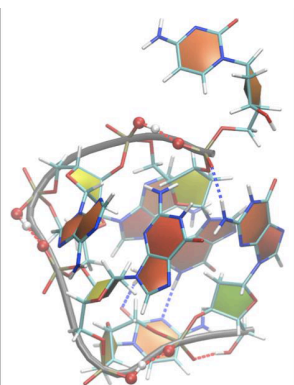
Thermodynamic integration,
metadynamics

Amber Force Field, Gromos96 Code

Laio et al, *JCP* 2002

Dal Peraro et al. *Curr Op. Str. Biol.* 2007

Mass spectrometry
(with Modesto Orozco)



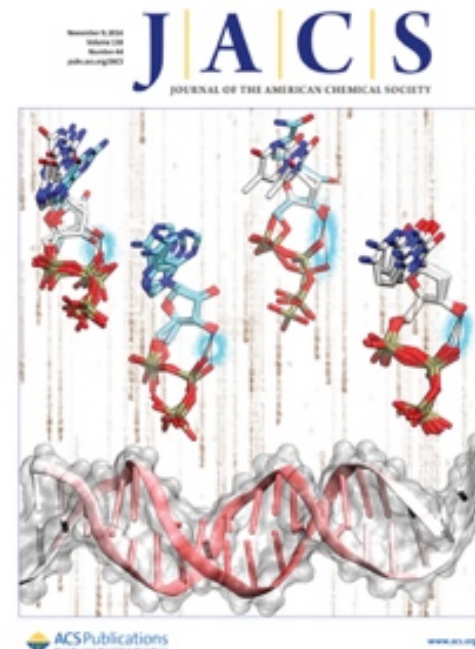
Arcella et al, *Angew. Chem.* 2015

Li et al, *J. Phys Chem Lett* 2017

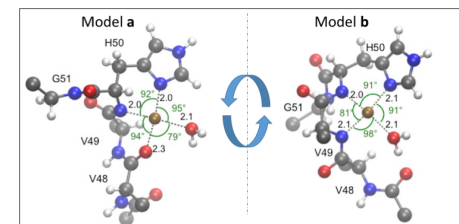
DNA pol enzymatic
reaction

Genna et al., 2016

Spotlights of JACS
publications
2016, 138, 14507



Copper(II)
binding sites

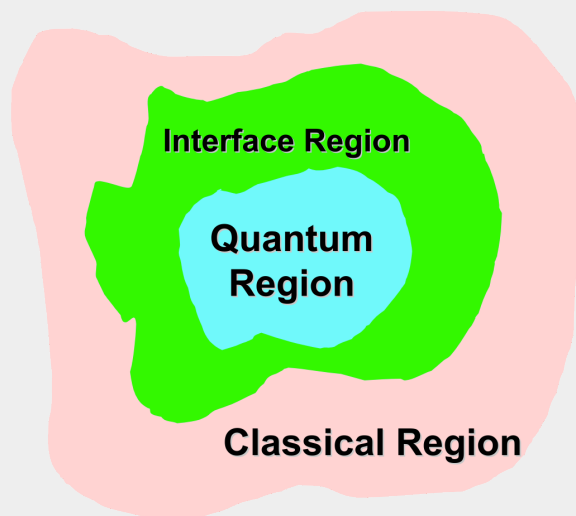


Villar-Pique et al., *PNAS*, 2016

Hybrid QM/MM molecular dynamics

$$L_{\text{CPMD}} = L_{\text{MD}} + \frac{1}{2} \sum_i \mu_i \int d\mathbf{r} |\dot{\psi}_i|^2 - V_{\text{DFT}}(\mathbf{R}, \mathbf{r}) + \sum_i \sum_j \Lambda_{ij} \left(\int \psi_i^* \psi_j d\mathbf{r} - \delta_{ij} \right)$$

$$\rho(\mathbf{r}) = \sum_i |\psi_i(\mathbf{r})|^2$$



$$V_{\text{DFT}}(\mathbf{R}, \mathbf{r}) = \sum_I \sum_{J>I} \frac{q_I q_J}{R_{IJ}} - \frac{1}{2} \sum_i \int d\mathbf{r} \psi_i^*(\mathbf{r}) \nabla^2 \psi_i(\mathbf{r}) + \int d\mathbf{r} V_{\text{N}}(\mathbf{R}) \rho(\mathbf{r}) + \frac{1}{2} \int d\mathbf{r} d\mathbf{r}' \frac{\rho(\mathbf{r}) \rho(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} + E_{\text{XC}}[\rho(\mathbf{r})]$$

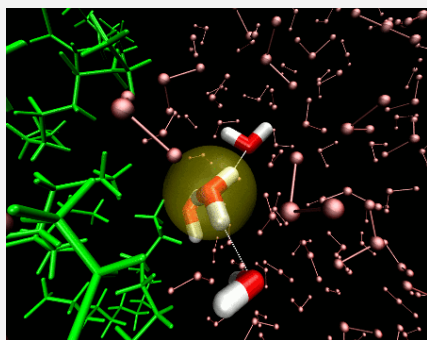
$$L_{\text{QM/MM}} = L_{\text{CPMD}} + L_{\text{MM}} - \sum_{i \in \text{MM}} q_i \int d\mathbf{R} \rho(\mathbf{r}) V_i(|\mathbf{R} - \mathbf{R}_i|)$$

QM/MM Simulations

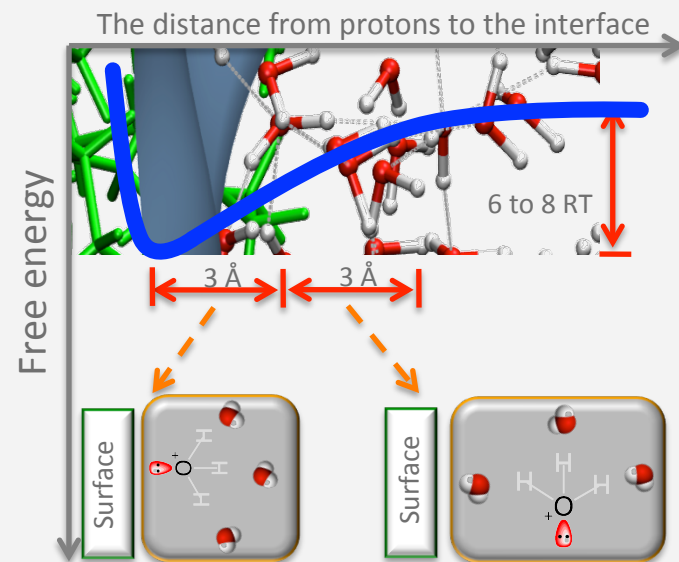
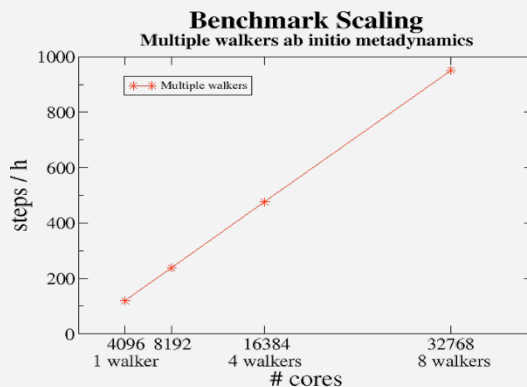
Problems:

- Poor scalability –sampling issues

CPMD calculation: Energetics of proton translocation



Zhang et al., *PNAS*, 2012



QM/MM Simulations

Problems:

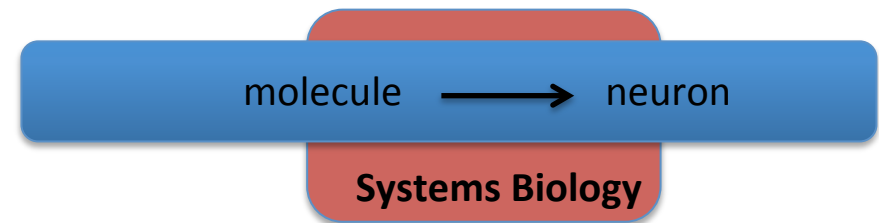
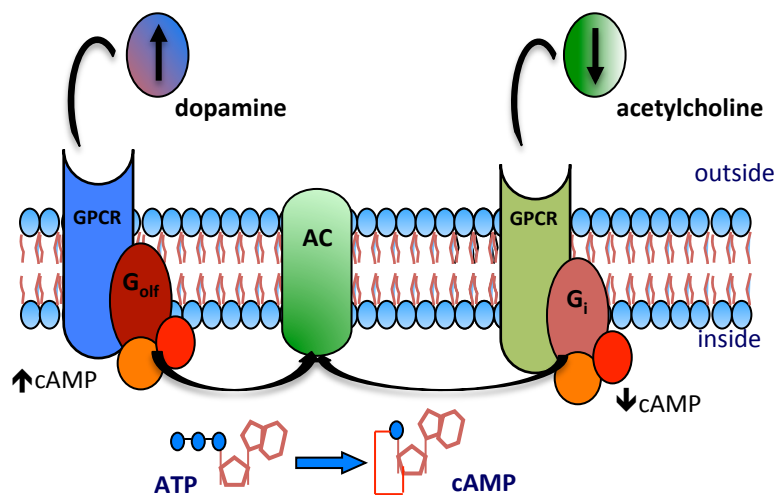
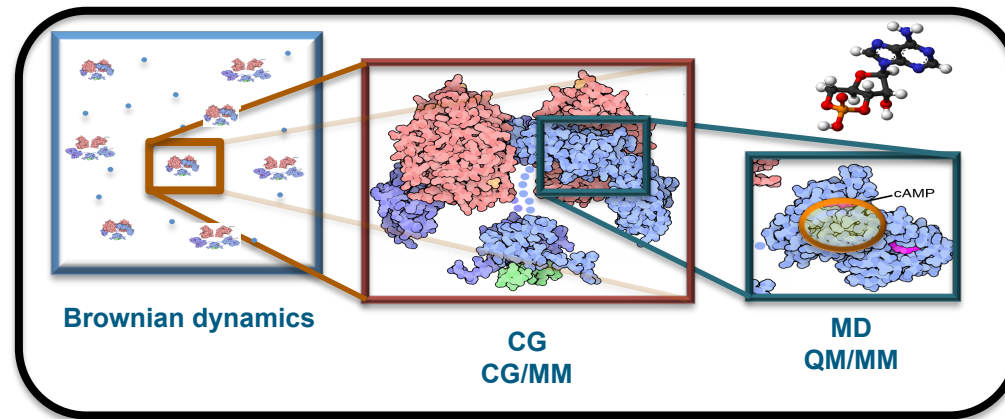
- Poor scalability –sampling issues
- Limited number of available force-fields
- Proprietary GROMOS96 license

New HPC-based Approach

- ✓ Design the architecture of an interface
- ✓ Design the protocol for data movement
- ✓ Develop the communication library
- ✓ Develop the QM/MM interface
- ⚠ Coupling to GROMACS (<http://www.gromacs.org/>)
- ✓ Writing the contributor manual
- ✓ Going open-source
- ✓ Performance and scaling optimizations
- ✓ Adopting different model for electrostatics treatment

- Viacheslav Bolnykh
RWTH-Aachen
- Emiliano Ippoliti
FZJ, Juelich
- Dr. Jógvan Magnus Hausgaard Olsen
University of Southern Denmark
- Dr. Simone Meloni
University La Sapienza
- Prof. Ursula Röthlisberger
EPFL, Lausanne
- Dr. Teodoro Laino
IBM, Zürich
- Dr. Valery Weber
IBM, Zürich
- Dr. Alessandro Curioni
IBM, Zürich
- Erik Lindahl
KTH, Stockholm

Multiscale simulations of neuronal receptors



Acknowledgements

Vania Calandrini, Alejandro Giorgetti, Massimo Sandal, Xevi Biarnes, Michael Leguebe, Giulia Rossetti, Mercedes Alfonso Prieto, Luca Maggi, Xiaojing Cong, Fabrizio Fierro
Prof Meyerhof's group (German institute of Nutrition, Potsdam, Germany),
Prof. Katrin Amunts (FZJ, Germany)
Prof. Sven Cichon (FZJ)
Prof. Andreas Bauer's (FZJ)
Prof. B. Neumeier (FZJ)

Funding: DFG, BMBF, Ernesto Illy Foundation, Human Brain Project, JARA, BioExel, HPC Leap



, May 2017

Neurodegenerative diseases

Gradual and progressive loss of neural cells, leading to nervous system dysfunction

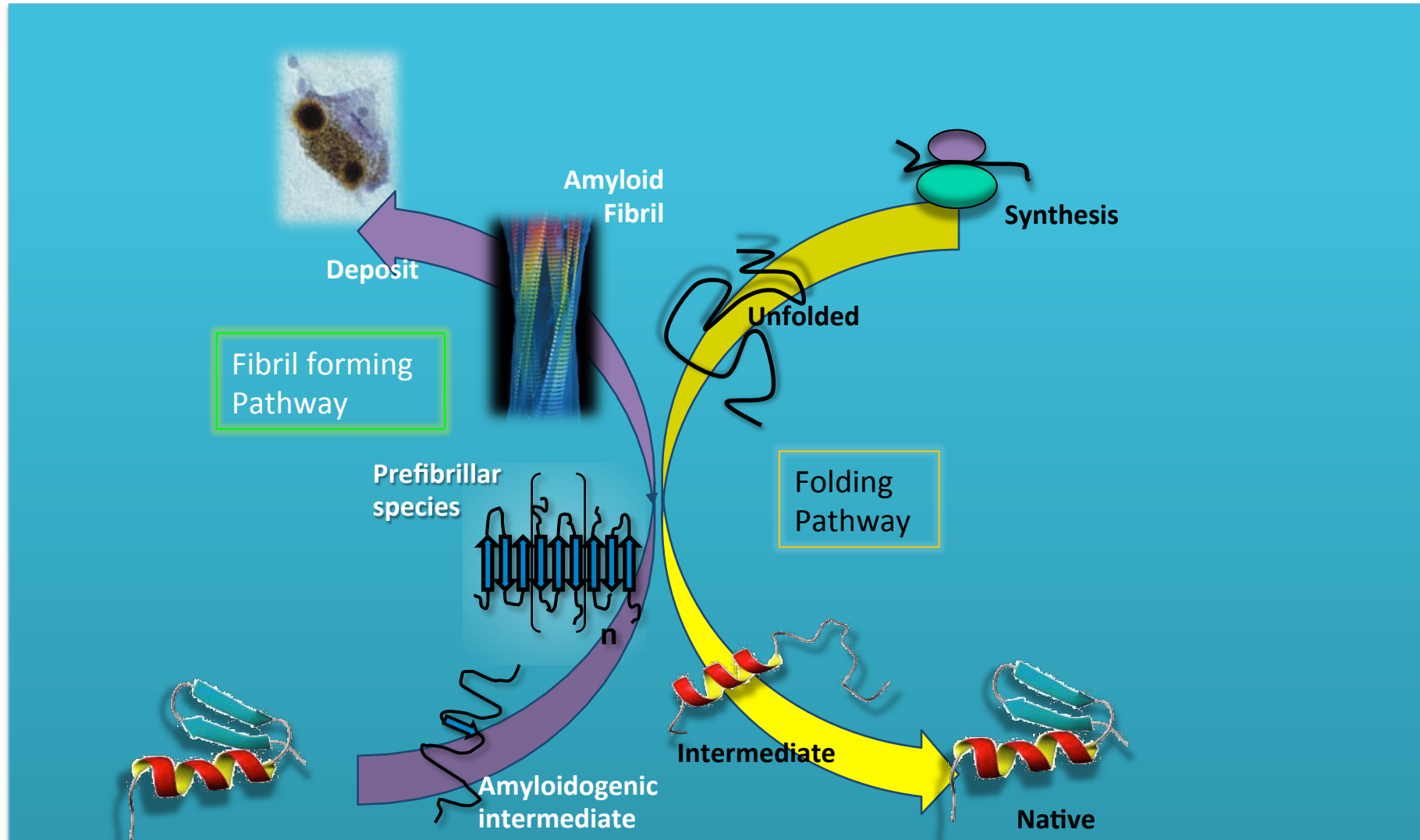
About 600 NDs, 50 million US individuals affected each year.

\$100 billion per year is spent on Alzheimer disease (AD) alone (2005) - immense emotional burden on patients and their relatives

As the number of elderly citizens increases, these costs to society also will increase.

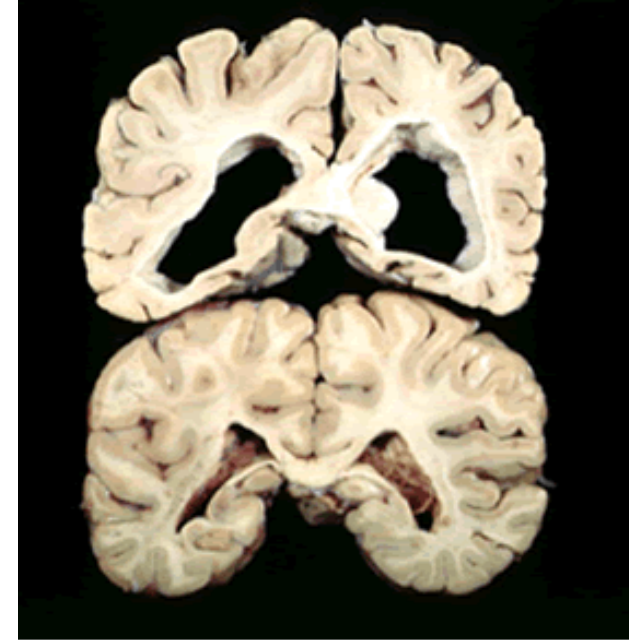
Brown et al, Environ Health Perspect. 2005;113, 1250

Protein fibril formation: Hallmark of neurodegenerative diseases



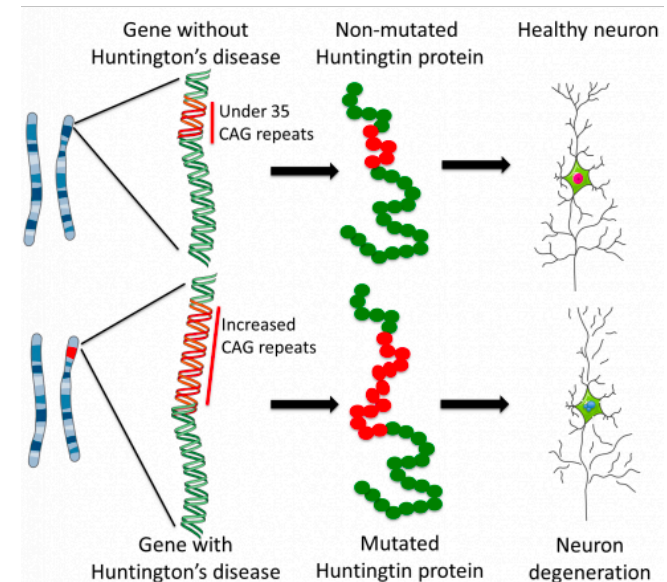
Huntington's disease

- Inherited fatal neurodegenerative disorder
- Over 10,000 people in the western world
- Uncontrolled mood swings
- No treatment



The human brain, showing the impact of HD on brain structure in the basal ganglia region of a person with HD (top) and a normal brain (bottom).

<http://kobiljak.msu.edu>



RNA as a target?

Huntington's disease

↓
CAG repeats

↓
Elongated RNA

↓
Pathogenic RNA_{CAG}/
protein complex

Inhibition

Activation



K_b (nM)
60 ± 30



K_b (nM)
700 ± 80



(r(5'-G₁GCAGCAGCC₁₀)₂)

Krauβ, Grieshe, Jastrzebska, Chen,
Rutschow, Achmuller, Dorn, Boesch,
Augsburger, Wanker, Schneider, Schweiger
(2013) *Nat. Comm.* 4:1511

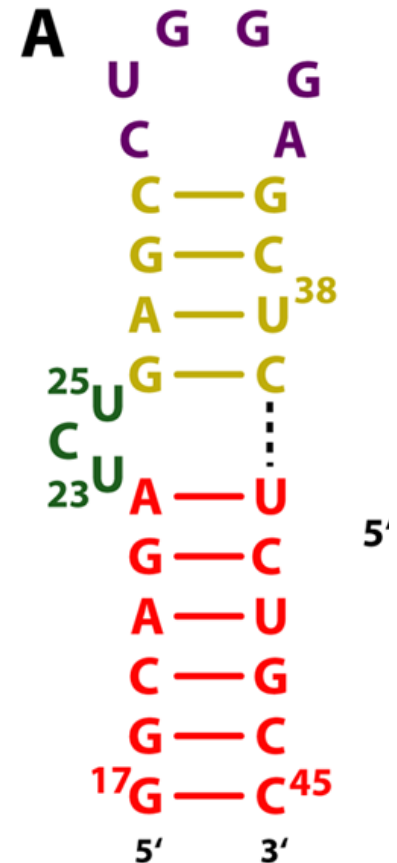
Kumar *et al.* (2012) *ACS Chem Biol.*
24:496.rna

Kiliszek, *et al.* *Nucleic Acids Res.* **2010**, 38 , 8370

Test system:

HIV-1 transactivation responsive RNA (TAR)

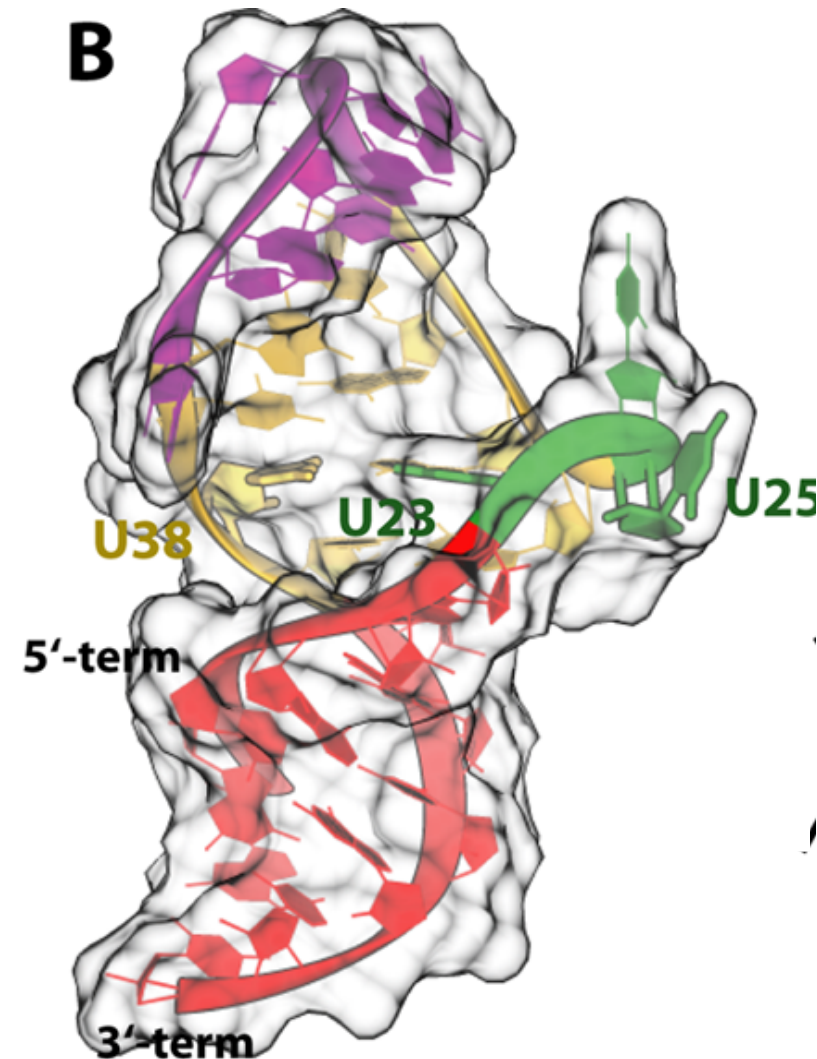
- 59 nucleobase TAR activates elongation of transcription of the integrated virus by forming a complex with the virally encoded HIV-1 Tat protein and with human cyclin T1



- Frankel Curr. Opin. Genet. Dev. 1992, 2, 293.
Weeks et al., Science 1990, 249, 1281.
Churcher, et al. J. Am. Chem. Soc. 2010, 132, 17643.
Wang et al. Med. Chem. 2009, 9, 379.
Kikuta, et al. E. J. Am. Chem. Soc. 2001, 123, 7911.
Sztuba-Solinska et al. J. Am. Chem. Soc. 2014, 136, 8402.

Structure: The Bulge

- NMR shows that TAR binds to Tat mimics and small ligands mostly through its “bulge” separating two helical regions (“upper” and “lower” stems). Partial stacking of U23 on A22 and C24 on U23 in the bulge generates a kink between the two stems



Aboul-ela et al. J. Mol. Biol. 1995, 253, 313,
 Aboul-ela et al Nucleic Acids Res. 1996, 24, 3974.
 Faber et al. J. Biol. Chem. 2000, 275, 20660.
 Du et al. Chem. Biol. 2002, 9, 707.
 Murchie et al. J. Mol. Biol. 2004, 336, 625.
 Davis et al. J. Mol. Biol. 2004, 336, 343.
 Davidson et al. Proc. Natl. Acad. Sci. U. S. A. 2009, 106, 11931, Nucleic Acids Res. 2011, 39, 248
 J. Mol. Biol. 2011, 410, 984.

July 3, 2017

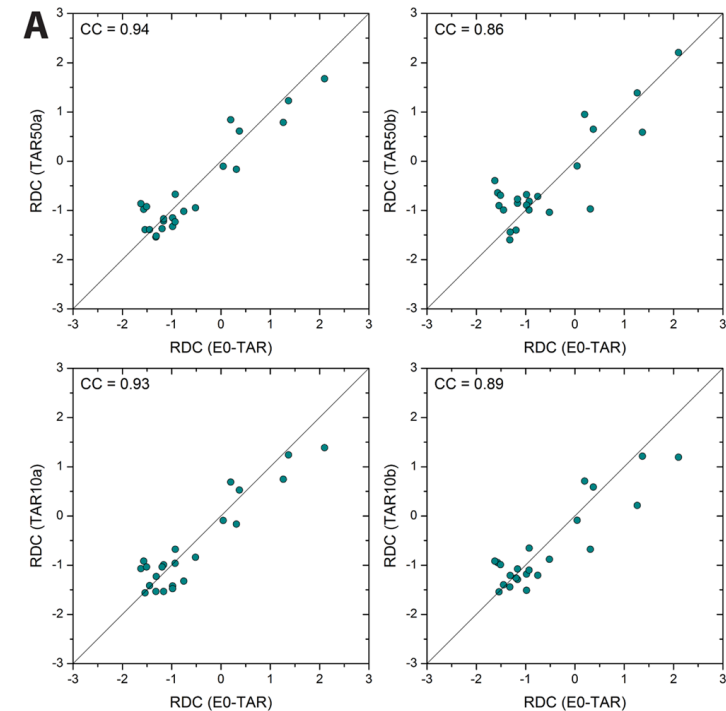
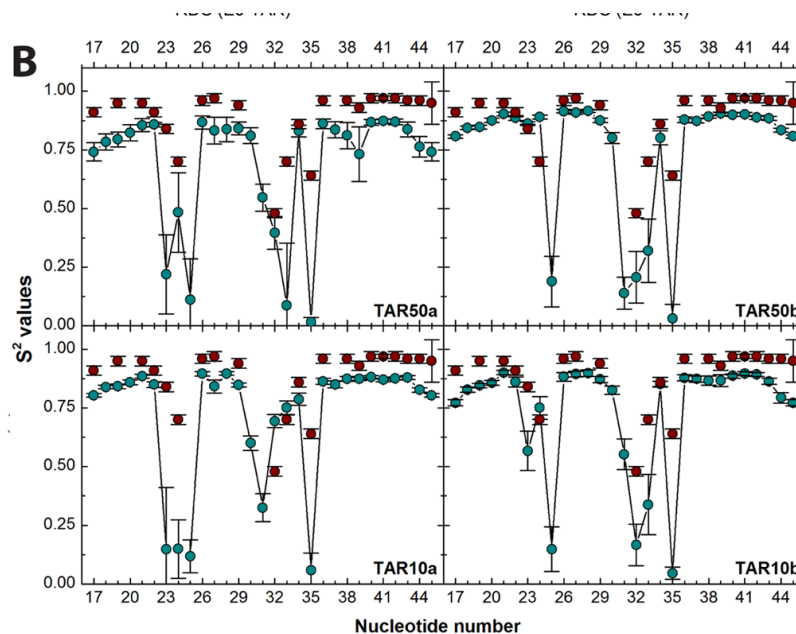
Computational Biomedicine

50

Investigating TAR internal motion



- Several μ s-long NPT AMBER-based MD simulations starting from *apo* and *holo* TAR give similar results



Perez, et al. Biophys. J 2007, 92, 3817.

Zhang et al. J. Am. Chem. Soc. 2003, 125, 10530.

Musiani et al, JACS 2014, 136, 15631

Free energy as a function of selected collected variables

1. Distance between the centers of mass of the ligand and of the CAG tract in RNA₁₀
2. Number of H-bonds between the ligand and the RNA
3. Number of hydrophobic contacts between the ligand and the RNA

Step I: Binding of known ligands

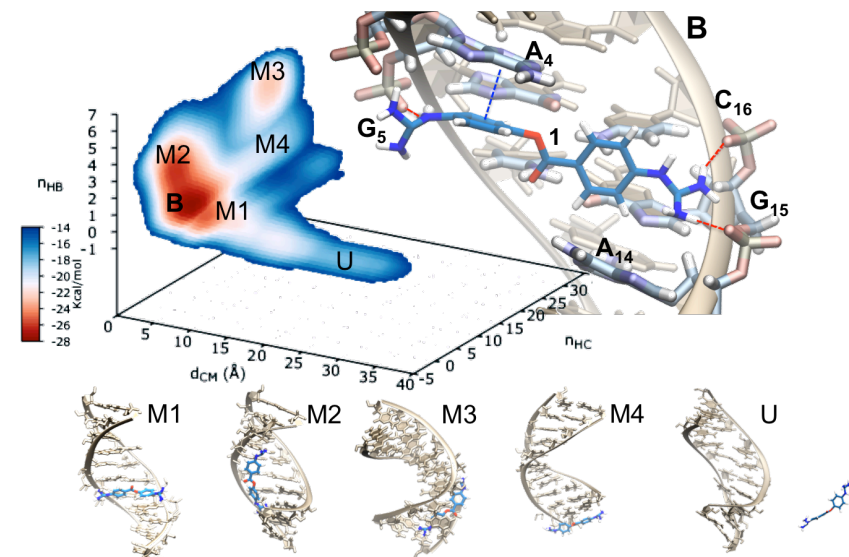
Huntington's disease:
Repeat expansion-based disease

CAG repeats fold into
expanded hairpins
(disease threshold: > 37 repeats)

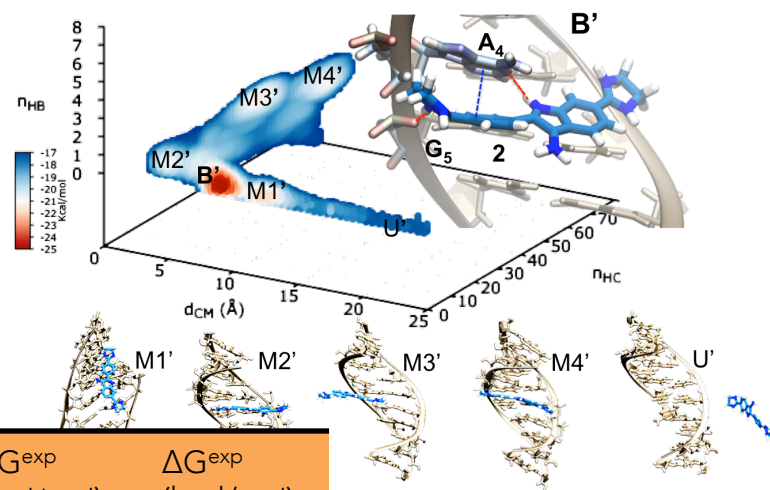
↓
Pathogenic
RNA_{CAG}/protein (e.g. MID1)
complex

↓
Increased translation
of neurotoxic HTT protein

Derive small
molecules to target
(CAG)_n



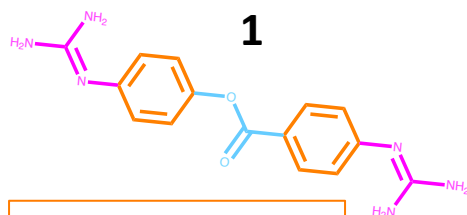
MD+ Well-
Tempered
Metadynamics



	ΔG^{exp} (kcal/mol)	ΔG^{exp} (kcal/mol)
1	-9.8 ± 0.7	~ -10
2	-8.4 ± 0.4	~ -8

Step II: Identification of new compounds

Scifinder
database search

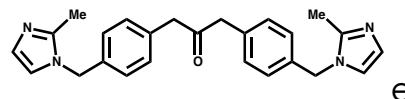
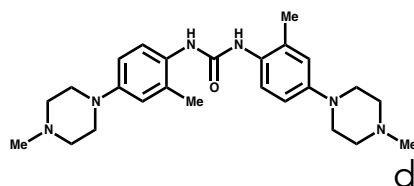
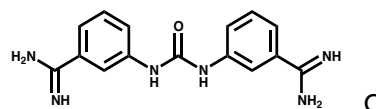
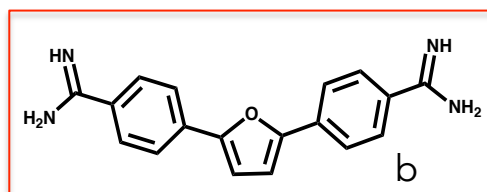
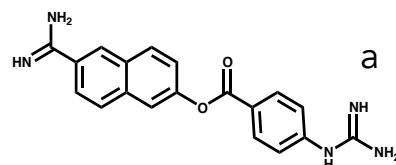


Two aromatic rings
interact with
Adenines base

Protonable
moieties

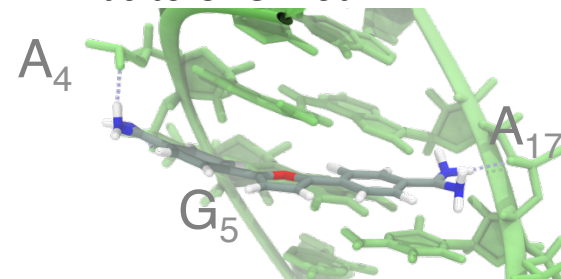
Linker: e.g. amide,
urea, furan ketone

25 compounds
selected



5 *In vitro/in cell*
essays

✓ Binds to CAG motif



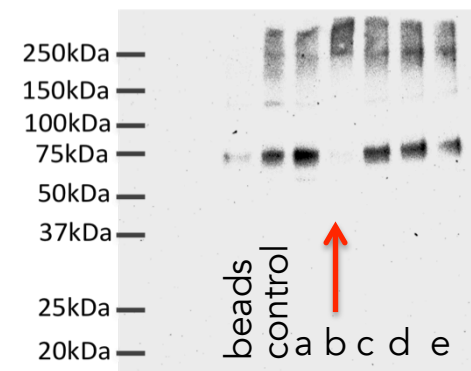
$$\Delta G^{\text{calc}} \sim 7 \text{ kcal/mol (CAG)}_2$$

$$\Delta G^{\text{MST}} = 5.7 \pm 0.2 \text{ kcal/mol (CAG)}_{18}$$

$$\Delta G^{\text{MST}} = 5.1 \pm 0.1 \text{ kcal/mol (CAG)}_{40}$$

✓ Inhibits the binding of
proteins (e.g. MID1) to
(CAG)*n* *in vitro* and in HD
cell models

(CAG)₇₂ pathogenic length



MID1-FLAG

Acknowledgements

Vania Calandrini, Alejandro Giorgetti, Massimo Sandal, Xevi Biarnes, Michael Leguebe, Giulia Rossetti, Mercedes Alfonso Prieto, Luca Maggi, Xiaojing Cong, Fabrizio Fierro, Anna Bochicchio

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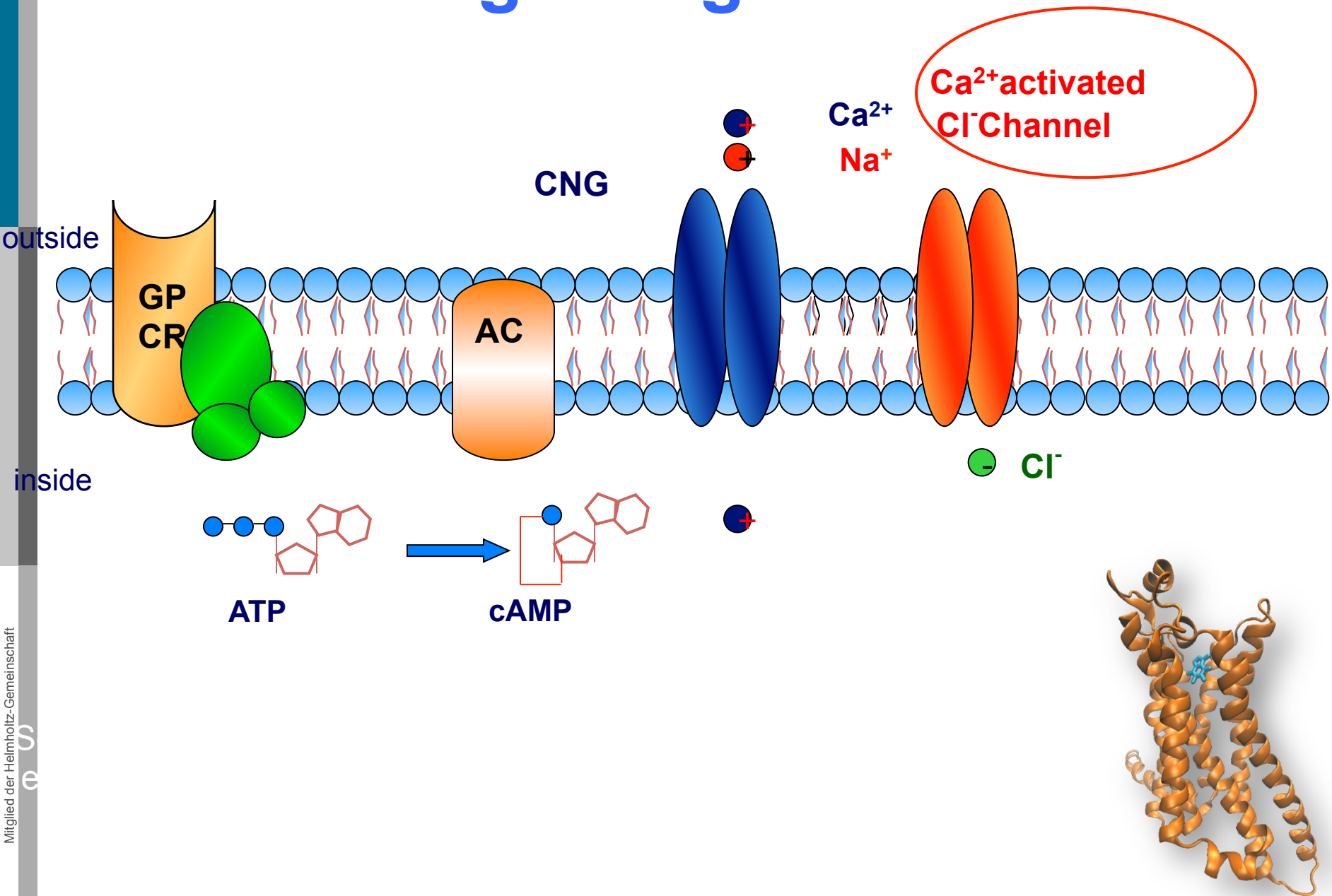
Prof. Sybille Krauss (DNZE, Bonn)

Funding: DFG, BMBF, Ernesto Illy Foundation, Human Brain Project, JARA, BioExel, HPC Leap



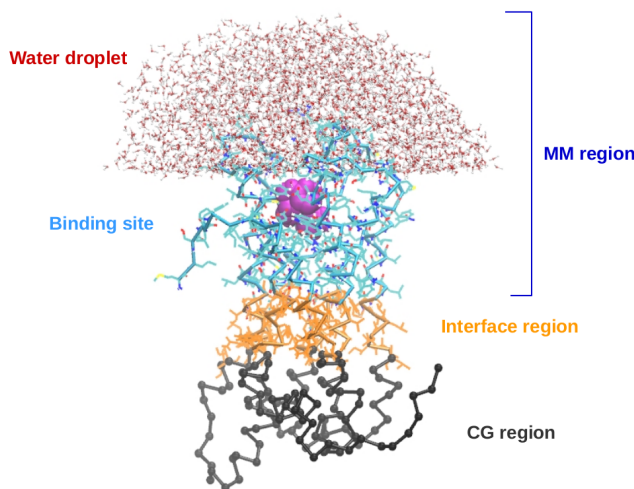
, May 2017

GPCR's signaling cascade



A multiscale approach for drug-affinity predictions

Molecular Mechanics/Coarse-grained approach

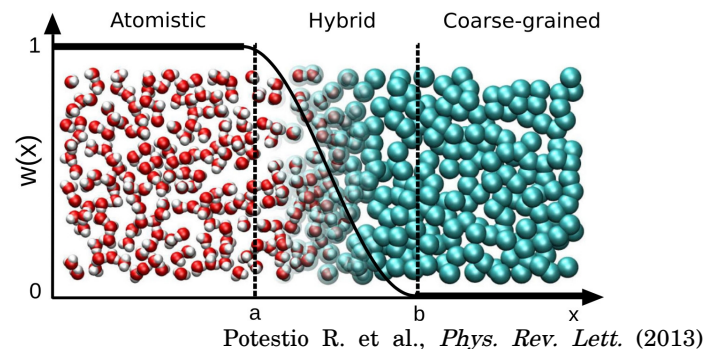


Leguebe M. et al., *PLOS ONE* (2012)

Hamiltonian-Adaptive resolution scheme

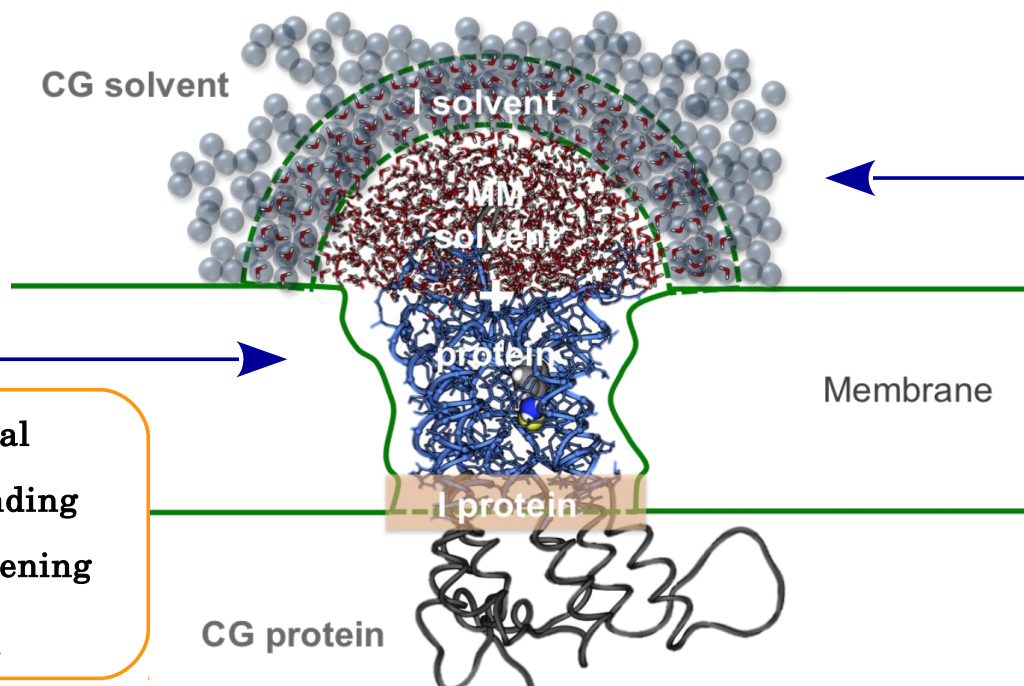
$$H = K + V^{p/p-w} + \sum_{\alpha}^N \left\{ \lambda_{\alpha} V_{\alpha}^{MM} + (1 - \lambda_{\alpha}) V_{\alpha}^{CG} \right\} - \sum_{\alpha}^N \Delta H(\lambda_{\alpha}).$$

$$\Delta H(\lambda_{\alpha}) = \frac{\Delta F(\lambda_{\alpha})}{N} + \frac{\Delta p(\lambda_{\alpha})}{\rho_0} \equiv \Delta \mu(\lambda_{\alpha}) = \frac{\Delta G(\lambda_{\alpha})}{N}.$$



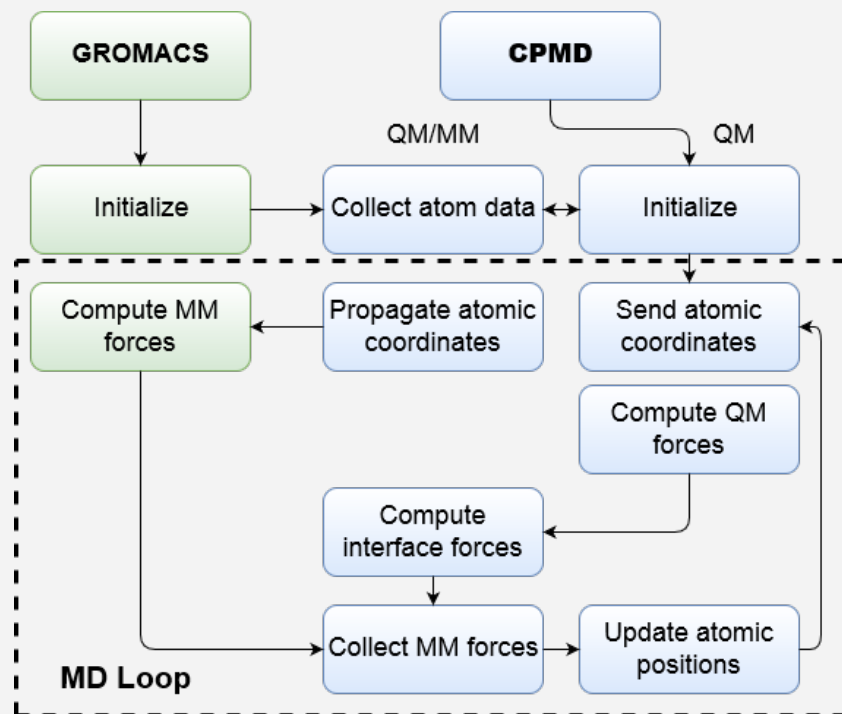
Dual-resolution
protein

Dual-resolution
solvent



Simulation of a **grand canonical ensemble** for calculation of **binding free energies** and **ligand screening** (need for **HPC resources**).

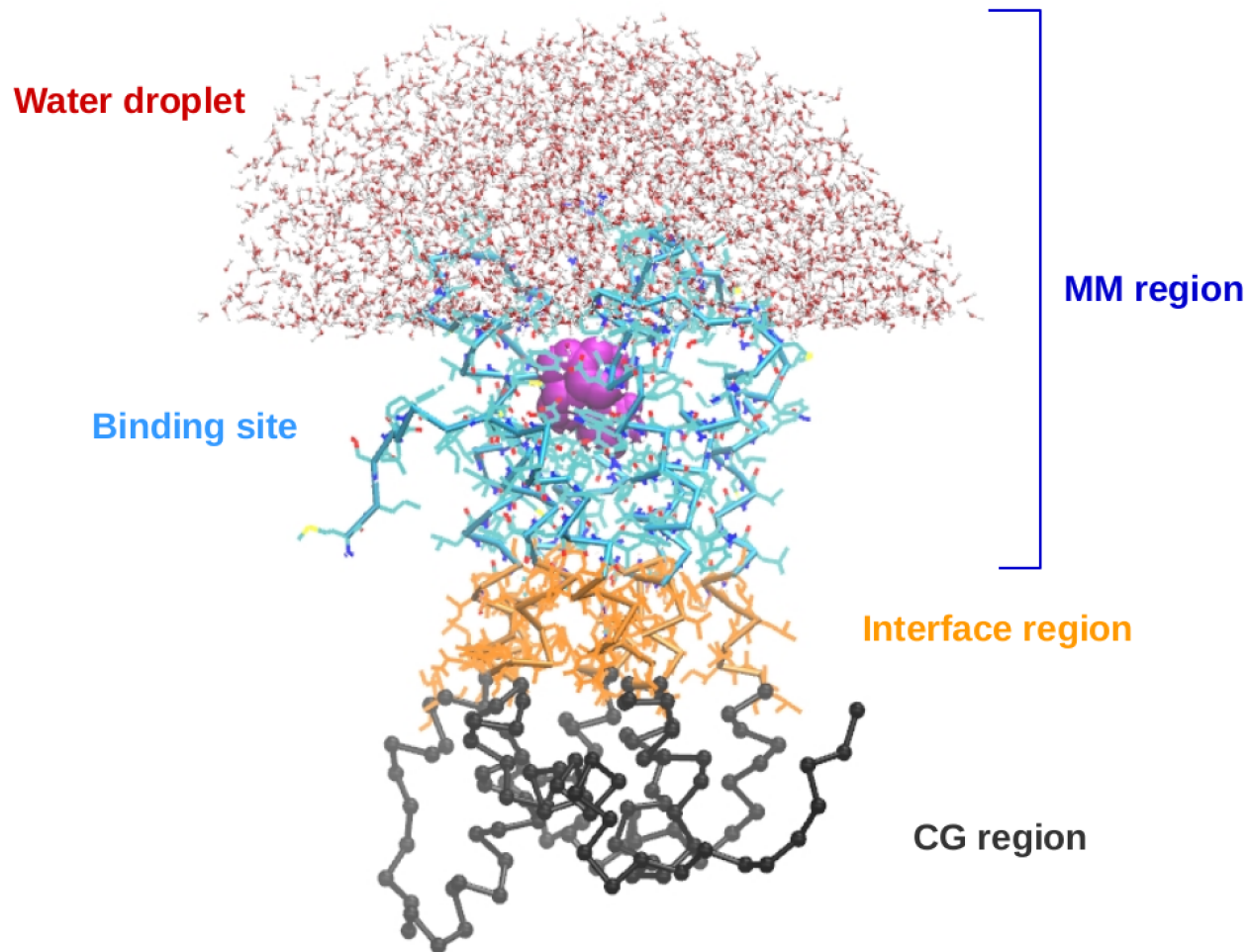
New HPC-based Approach



- Multiple-Program Multiple-Data approach using the ad-hoc communication library
- Fully exploits the efficient parallel architecture of both CPMD and the MM code
- Allows coupling to virtually any MM code

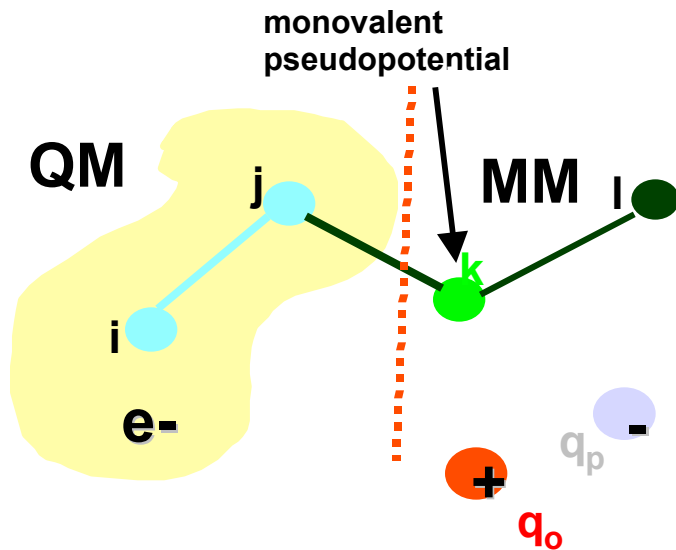
Improving the scaling performance allows getting better sampling

Molecular Mechanics/Coarse-grained approach



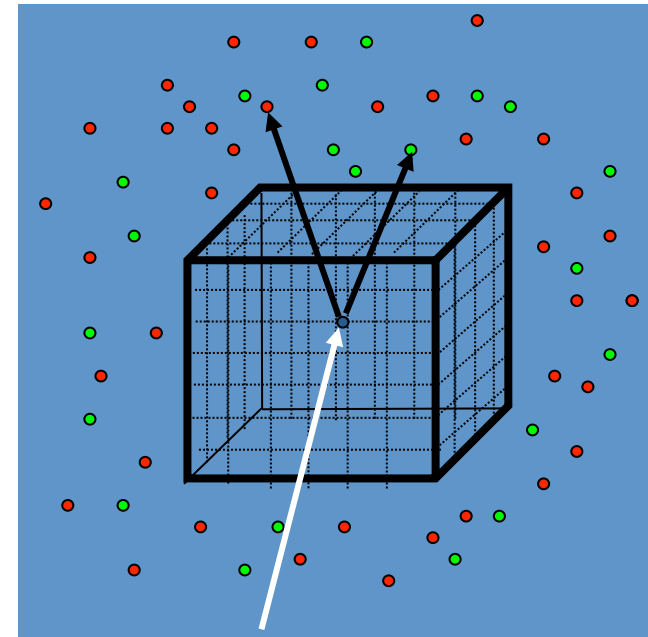
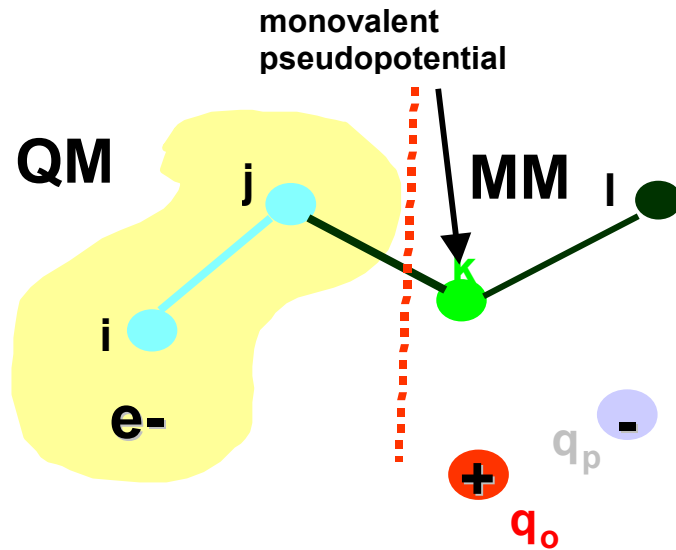
Leguebe M. et al., *PLOS ONE* (2012)

- Efficient sampling of side chain at binding site
- Simplified representation of the rest



MM atoms/QM atoms bonds: monovalent pseudopotentials

Angle bending and dihedral distortions: Classical force field

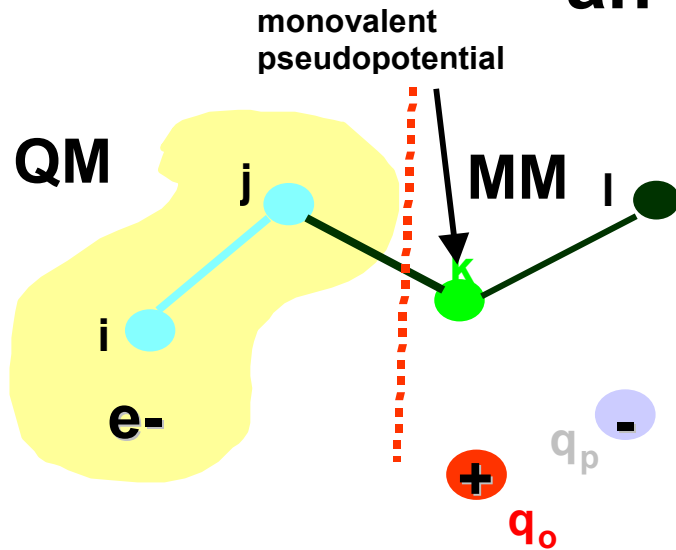


$$E_{QM/MM}^{st} = \sum_{I \in MM} q_I \int dr \rho(r) / |r - R_I| + \sum_{\substack{I \in MM \\ J \in QM}} v_{vdw}(R_{IJ})$$

1-Electron density is overpolarized at short range: *electron spill-out* problem

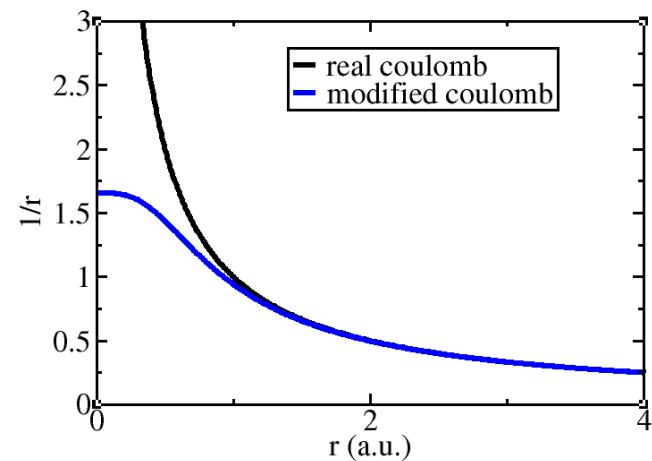
2- # operations $\sim N_{rsgrid} \times N_{MM} \sim 1,000 \times 10,000$

1-Spill out: Replacing the Coulomb potential with an *ad hoc* function



$$E_{QM/MM}^{ele} = \sum_{i \in MM} q_i \int dr \rho(r) v_i(|r - r_i|)$$

$$v_J(r) = \frac{R_{cJ}^4 - r^4}{R_{cJ}^5 - r^5}$$



R_{cJ} =cutoff radii, tested in Laio et al JCP 2002

each node with descendants represents the inferred most recent common ancestor of the descendants, and the edge lengths in some trees may be interpreted as time estimates. Each node is called a taxonomic unit. Internal nodes are generally called hypothetical taxonomic units, as they cannot be directly observed.