### ADME models in the drug discovery pipeline



Drug Discovery Today

#### **eADMET** prediction



#### ADME models (I)

Following models are useful for *in silico* design:

primary models

solubility intestinal absorption bioavailability metabolic stability blood-brain-barrier permeation mutagenicity cardial toxicity (*h*ERG-channel) plasma protein binding secondary models

transport (uptake and efflux) common toxicity hepatotoxicity (PXR, CAR) nephrotoxicity immunotoxicity neurotoxicity (receptor binding) drug-drug interactions (Cytochrome P450)

Covered in this lecture and the upcomming lectures

### ADME models (II)



#### Why is ADME prediction that important ?



Reasons that lead to failure or withdrawl of a potential drug around 1995 – 2000

### Why is ADME prediction that important ? (II)

Our aim is to reckognize unsuitable compounds as soon as possible:

- saving resources
- avoiding unnecessary clinical trials
- The later a drug has to be withdrawn, the more expensive it gets.

"Fail early, fail fast, fail cheap"



### Compound selection for the High Throughput Screening (HTS)

Project virtual library of 100,000 members



#### solvation versus solubility



### Solubility models (I)

Direct computation of the solubility from a thermodynamic cycle (lattice energy, heat of solvation) is possible, but

- 1. The prediction of the lattice energy by computational methods requires knowing the space group of the crystal
- 2. Computation of the heat of solvation is errorprone itself

Thus, mainly QSAR approaches are applied.

However, automated workflows that derive specifically tailored force fields for the molecule at hand exist, although these are computationally very demanding.



Lit: Neumann MA et al. *Nat. Commun* **6** (2015) 7793. Reilly et al. *Acta Cryst. B* **72** Pt4 (2016) 439-459.

### Solubility models (II)

descriptors: connectivity indices



r<sup>2</sup>=0.89, q<sup>2</sup>= 0.84, se = 0.98, n=120, F=297.80

Lit. C. Zhong et al. J. Pharm. Sci. 92 (2003) 2284

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### Solubility models (III)

Further approaches show that the applied descriptors must account for lipophilic and H-bond properties, as well as the flexibility of the compounds

Lit: A. Cheng et al. *J.Med.Chem.* **46** (2003) 3572

D. Butina et al. J.Chem.Inf.Comput.Sci. 43 (2003) 837

Besides common QSAR equations, also machine learning approaches are used:

Lit: A. Yan et al. J.Chem.Inf.Comput.Sci. 43 (2003) 429

J.K. Wegener et al. Ibid 43 (2003) 1077

Z. Ye, D. Ouyang J. Cheminf. 13 (2021) 98

S. Lee et al. ACS Omega 7 (2022) 12268

Predicting the solubility of a compound in various solvents and solvent mixtures is likewise an issue for chemical synthesis.

### Absorption

How much and how fast is a substance absorbed?

Drugs should be orally applicable for convenience After passing the stomach, they are resorbed from the colon/intestine into the blood. Transport via the portal vein into the liver.



### Absorption in the duodenum (I)

Uptake of a substance into the systemic circulation



### Absorption in the duodenum (II)

Uptake of a substance into the systemic circulation





- A transcellular (passive diffusion)
- B paracellular
- C active transport
- D transcytosis

# Cross-section from the colon wall

### Absorption in the duodenum (III)



phospholipid



De Groot et al. Science 294 (2001) 2353 7th lecture Modern Methods in Drug Discovery WS22/23

### Caco-2 cell monolayer

Experimental approach for the prediction of intestinal absorption

monolayer of a culture of cells that are derived from a colon cancer



Advantage: reproducable results, in good agreement with *in vivo* studies

Disadvantage: these cells exhibit somewhat different metabolic properties than cells for the duodenum (MDR1 transporter = P-glycoprotein is over expressed, which leads to increased efflux of many substances)

Besides Caco-2 cells, also synthetic membranes are used for screening

# What factors determine the passive diffusion through lipidbilayers ?

Small molecules should pass through faster than large descriptor: molecular weight (MW) and molecular shape

phospholipid bilayers are lipophilic on the inside Thus, lipophilic molecules should pass through the interior faster descriptor: logP (water/*n*-octanol partition coefficient)

phospholipid bilayers have a hydrophilic surface descriptors: number of H-bond donors and acceptors observation: the permeability is related to the heat of solvation

#### Descriptors derived from molecules to predict ADME properties

Frequently occuring in QSAR equations and machine learning algorithms are:

logP water/*n*-octanol partition coefficient number of violations of Lipinski's rule of five topological indices (e.g. from fingerprints) polar surface area count of polar atoms (can form hydrogen-bonds) count of titratable groups (acids, bases) count of halogen atoms (lipophilic)

Lit: S. Lee et al. ACS Omega 7 (2022) 12268

#### Lipinski's Rule of 5

Combination of descriptors to estimate intestinal absorption. Insufficient uptake of compounds, if

Molecular weight > 500

logP > 5.0

slow diffusion too lipophilic

> 5 H-bond donors (OH and NH)

too many H-bonds with the

>10 H-bond acceptors (N and O atoms) head groups of the membrane



C.A. Lipinski et al. *Adv. Drug. Delivery Reviews* **23** (1997) 3. 7th lecture Modern Methods in Drug Discovery WS22/23

### **Polar Surface Area (PSA)**

The PSA is defined as the part of the molecular surface of a compound that stems from the nitrogen and oxygen atoms, as well as the polar hydrogens bonded to them.

= a quantitative measure for the ability to form H-bonds



Like all other 3D descriptors the PSA is in general dependent from the conformation.

### **Models for absorption**

New studies show, however, that there is a sound correlation between Caco-2 absorption and uptake (fractional absorption) in human (%FA) regardless of possible conformers.



Lit: D.E. Clark, *J.Pharm.Sci.* **8** (1999) 807; *Drug Discovery Today* **5** (2000) 49; K. Palm et al. *J.Med.Chem.* **41** (1998) 5382

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#### Pharmacokinetic and Bioavailability

The body/organism is regarded as an open system that tries to restore the equilibrium after each disturbance/dosage



The body is partitioned into a series of compartments. Between these compartments there is a constant flow / exchange.

### distribution / invasion

The total path of a substance can be separated into

- 1) diffusion in the solvent
- 2) diffusion through tissue and membranes
- 3) transport by the blood
- 4) a) diffusion to the receptorsb) diffusion into other comparts
  - b) diffusion into other compartments
  - c) diffusion into elimination organs
- 5) irreversible elimination

invasion (according to Dost) ≈ distribution

Ibsorption

High constant of elimination: short period anesthetics Low constant of elimination: antibiotics

### Volume of distribution and dosage



### Invasion / systemic exposure

The full concentration can only be achieved by intravenous application. Otherwise invasion and elimination interact. This correspond physicochemically to subsequent reaction.



## The principle of Dost (I)

Dependence of the concentration profile for different dosage



Between two sample points, the area S (transit) below the curve can be obtained by integration of the Bateman function as:



Total clearance: volume that is cleared per unit of time

 $Cl_{tot} = \frac{\ln 2}{t_0} V [\text{volume/time}]$ 

Corresponding areas correspond to the ratio of the doses

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## The principle of Dost (II)

The reference curve is obtained by intravenous application of the dose





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#### Experimental data for pharmacokinetic models

#### chemical data

partition coefficients metabolic turnover rates

V<sub>max</sub>, K<sub>m</sub>, K<sub>i</sub> solubility vapour pressure diffusion constant protein binding constants

#### biological data

anatomic dimensions flow of blood through the organs volume of organs

respiration body mass

age, gender extent of physical activity

### Pharmacokinetic models (I)

#### **Compartment models**

assumption:

no metabolic conversion inside the compartments



The concentration profile with time can be calculated by using linear differential equations

Lit: J.E.W.Beneken, J.H.van Oostrom "Modeling in Anethesia", *Journal of Clinical Monitoring and Computing* **14** (1998) 57-67.

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#### Pharmacokinetic models (II)

Systemic blood circulation as electric network (1930)



Simulation via analog computers (patch cords between the modules, resistors, capacitors)

applicability: inhalative anesthetics (low metabolic conversion, lipophilic, are exhaled)

Lit: G.Fleischli, E.N.Cohen Anesthiology 27 (1966) issue 1, 64-69.

### Distribution

From within the plasma the drug has to reach other compartments, depending on its target.

Substances that act on the central nervous system (CNS) have to cross the blood-brain barrier. Conversely, other drugs should not pass this barrier.



Besides passive diffusion, active transport has to be considered. Charged and polar substance require active transport.

### Plasma protein binding / Distribution

The available concentration of drugs can be reduced due to binding to other proteins. This occurs in the plasma, the extracellular and interstitial fluid.

$$A+B^{\rightarrow}AB$$
 with  $v_{bind} = k_{bind}[A][B]$   
AB $^{\rightarrow}A + B$  with  $v_{diss} = k_{diss}[AB]$ 

In the equilibrium state no change is measurable, thus

$$k_{bind}[A][B] = k_{diss}[AB]$$
$$K = \frac{[AB]}{[A][B]} = \frac{k_{bind}}{k_{diss}}$$

Binding proceeds according to the Langmuir's absorption isotherm (the heat of absorption is independend from the degree of coverage) and therefore fulfills the law of mass action [Massenwirkungsgesetz])

Besides proteins also mucopolysaccharides (binding- and supporting tissue (stroma)) can absorb substances.

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# Metabolism (I)

(bio-)chemical reactions of xenobiotics in the body

First pass effect:

Extensive metabolization of mainly lipophilic molecules, such with MW>500, or those that have a specific affinity to certain transporters, during the first passage through the liver

Phase I:

Oxidation, reduction and hydrolysis esp. Cytochrome P450 enzymes

Phase II:

Conjugation with small molecules (e.g. glutamine)

Phase III:

elimination by transporters



experimental (in vitro) methods:

human liver microsomes, hepatocytes and recombinant P450 enzymes (expressed in *E. coli, or yeast cells*)

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### **Elimination / Excretion**

Elimination comprises all processes that lead to removing of a substance from a compartment. These can also be of metabolic nature.

Lipophilic substances can be excreted using bile [Gallensaft], hydrophilic compounds via urine..

In general:

MW <300 300-500 >500

bile bile & urine urine



### **Elimination / Clearance**

Metabolic paths (overview)



## Elimination / Clearance (III)

From the physico-chemical point of view, elimination of a substance is a 1st order decay process (depending on the present concentration of the compound)

 $\vec{A} = B$  with v = k[A] = k rate constant of elimination



### The blood-brain barrier (BBB)

Cross section through a cappilary vessel



According to: J.-M. Scheerman in *Pharmacogenomics*, J.Licinio & Ma-Li Wong (Eds.) Wiley-VCH (2002) pp. 311-335.

#### Function of the blood-brain barrier

*in silico* prediction of the blood-brain barrier permeability in the course of pre-clinical development is particularly important, since

- only substances that shall act on the central nervous system (CNS), should pass the blood-brain barrier effectively.
- BBB-screening is particular "expensive" (testing on animals not avoidable: microdialysis, isotope labeling)
- models using artificial membranes (endothelial cells) are still in development.

### **Blood-Brain Barrier (II)**

As a measure for the permeability of the blood-brain barrier, the logarithmic ratio of the concentrations is used logBB = log([brain]/[blood]) range: -2.00 to +1.00 Mainly in the blood -1.0 < logBB < 0.3 mainly in the brain

It can be assumed that the logBB has been determined for about 300 drugs, only. However, for much more compounds a qualitative assignment (CNS+ or CNS–) is known.

Lit. D. E. Clark, J. Pharm. Sci. 8 (1999) 815

### **Blood-Brain Barrier (III)**

In contrast to the absorption from the duodenum, the polarity of the compounds that cannot be described by the PSA comes into account. Example:

	PSA	logBB	ClogP	polarizablity	(AM1)
benzene	0	-0.69	2.1	1	3.8
3-methylpentane	0	2.01	3.7	14.8	

An according QSPR equation was derived logBB = a PSA + b ClogP + c with r = 0.887

Lit. D. E. Clark, *J.Pharm.Sci.* **8** (1999) 815 F. Lombardo et al. *J.Med.Chem.* **39** (1996) 4750

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#### **Formerly used descriptors**

Each of these terms is correlated to logBB by itself:

- logP fragment based (MlogP, ClogP,...)
- Polar surface area contributions from N, O and H atoms
- hydrogen-bond donors and acceptors numerical count
- size and shape molecular volume and globularity

#### **Descriptors for size and shape**

Connected to the shape of the molecule are:

Molecular volume, globularity, number of rotatable bonds

globularity:

Ratio of the surface (assuming the molecule would be a perfect sphere) to the actual surface. Always < 1



Principle components of the molecular geometry:

3D extension of the molecule in space

#### New descriptors for size and shape

- Descriptors such as the globularity are correlated to the molecular weight and the number of hydrogen atoms

+ Replaced by three terms derived from the molecular geometry



#### **BBB-model with 12 descriptors**

Descriptors mainly from QM calculations: electrostatic surface, principal components of the geometry, H-bond properties



#### **ADME** – historical development

1960 Corwin HanschQSAR for small data setslogP for toxicity

- 1980 in vitro studies replace in vivo studies
- 1990 first *in silico* ADME (computer) models,docking into protein structures,homology modeling of proteins (CYP P450)
- 1997 Lipinski's rule of five for (passive) absorption
- 2003 X-ray structure of human CYP2C9 (10G2.pdb)
- 2004 X-ray structure of human CYP3A4 (1TQN.pdb)
- 2005 X-ray structure of human CYP2D6 (2F9Q.pdb)
- 2015 Routine use of machine learning for ADME property prediction

#### Web-based online tools

A number of institutes and companies have put up servers for the prediction of ADME related properties.

Usually these apply Java-applets that allow drawing molecules, allow input either as SMILES string or one of the may 3D coordinate files.

A summary including hyperlinks is offered by the *Virtual Laboratory* 

http://146.107.217.178/online.html

Lit. I.V. Tetko, *Mini Rev.Med.Chem.* **8** (2003) 809.

I.V. Tetko et al., J.Comput.-Aided Mol.Des. 19 (2005) 453.