Integration of biomarkers in risk assessment; a toxicological perspective

#### Rory B. Conolly, Sc.D.

Presented at the ENCIS Workshop "Integration of Biomarkers in Cancer Risk Assessment", Mitland Hotel Utrecht, Arienslaan 1, 3573 PT Utrecht, The Netherlands Thursday, October 19, 2006

### Outline

- Using mechanistic data to reduce uncertainty in risk assessment
- Formaldehyde nasal SCC in rats
- Mechanistic studies of the rat tumors
- Risk assessment driven by the data
- IARC



The problem



The solution

### Approach

Quantitative modeling of the mode or mechanism of action to predict doseresponse and time-course behaviors



(Formal + intuitive modeling)

### Biologically based computational models form natural bridges between research and risk assessment



### Reduction of uncertainty in risk assessment



### Dose-response assessment for formaldehyde





### Formaldehyde bioassay results



**Exposure Concentration (ppm)** 

### Normal respiratory epithelium in the rat nose



### Formaldehyde-exposed respiratory epithelium in the rat nose (10+ ppm)



### Dose-response for cell division rate







### DPX submodel – simulation of rhesus monkey data



#### DPX and direct mutation

• Direct mutation is assumed to be proportional to the amount of DPX:

 $mutation = KMU \cdot DPX$ 

- Low-dose linear!
  - Is KMU big or small?

Summary of biomarker dose-response inputs to the clonal growth model

- Cell replication
  - J-shaped
- DPX
  - Low dose linear

### CFD Simulation of Nasal Airflow (Kimbell et. al)





### Flux bins

 Nasal surface area partitioned into 20 bins ranked according to flux of formaldehyde predicted by the CFD model



### 2-Stage clonal growth model (MVK model)



### Calculation of the value of KMU

- Grid search
- Optimal value of KMU was zero
  - Modeling predicts that direct mutation is not a significant action of formaldehyde
- 95% upper confidence limit on KMU was estimated

### Maximum likelihood grid search



### Optimal value of KMU is zero



# Hockey stick model fit to raw data to cell division dose-response



ppm formaldehyde

ppm formaldehyde

### Simulation of tumor response in rats



### From rats to humans

# Computational fluid dynamics models of the nasal airways



### Human assessment



## Baseline calibration against human lung cancer data



### Human risk modeling



### Upper bound on KMU



### Final model: Hockey stick and 95% upper confidence limit on value of KMU



33

Predicted human cancer risks (hockey stick-shaped dose-response for cell replication; optimal value for KMU)



### "Negative risk" using raw dose-response for cell replication



### Lutz & Kopp-Schneider: Tumor incidence with J-shaped cell replication & linear mutation



36

## 1999 - 2004 CIIT Inhaled ppm CFD modeling Cell killing Cell proliferation Tissue dose Cancer model (Clonal growth) Mutagenicity (DPX) Tumor response 37



### Inhaled ppm



Cancer model (LMS)

Tumor response

### 1991 U.S. EPA



### Make conservative choices when faced with uncertainty

- Use hockey stick-shaped cell replication
- Use a 95% upper bound on the dose-response for the directly mutagenic mode of action
  - Statistically optimal model has 0 (zero) slope
- Risk model predicts low-dose linear risk.
- Optimal, data based model predicts negative risk at low doses

# Summary: CIIT assessment of formaldehyde cancer risk

- Either no additional risk or a much smaller level of risk than previous assessments
- Consistent with mechanistic database
  - Direct mutagenicity
  - Cell replication

### IARC 2004

- Classified 1A based on nasopharyngeal cancer
- Myeloid leukemia data suggestive but not sufficient
  - Concern about mechanism
  - British study negative
- Reclassification driven by epidemiology
- *In my opinion* inadequate consideration of regional dosimetry and mechanistic data from rat studies

### Whole nose







#### Anterior nose

#### nasopharynx

### Formaldehyde summary

- Using mechanistic data to reduce uncertainty in risk assessment
- Formaldehyde nasal SCC in rats
- Mechanistic studies of the rat tumors
- Risk assessment driven by the data
- IARC

### Reduction of uncertainty in risk assessment



### Disclaimer

EPA has sponsored Dr. Conolly's attendance at this meeting. This presentation is not a statement of official policy of the United States Environmental Protection Agency. Acknowledgements (I)

• Many, many investigators at CIIT (and elsewhere) who have studied formaldehyde.

### Acknowledgements (II)

- Colleagues who worked on the clonal growth risk assessment
  - Fred Miller, Julian Preston, Paul Schlosser, Julie Kimbell, Betsy Gross, Suresh Moolgavkar, Georg Luebeck, Derek Janszen, Mercedes Casanova, Henry Heck, John Overton, Steve Seilkop

### End