Benzene toxicokinetics in humans: exposure of bone marrow to metabolites

Karen H Watanabe, Frédéric Y Bois, Joan M Daisey, David M Auslander, Robert C Spear

Abstract

A three compartment physiologically based toxicokinetic model was fitted to human data on benzene disposition. Two separate groups of model parameter derivations were obtained, depending on which data sets were being fitted. The model was then used to simulate five environmental or occupational exposures. Predicted values of the total bone marrow exposure to benzene and cumulative quantity of metabolites produced by the bone marrow were generated for each scenario. The relation between cumulative quantity of metabolites produced by the bone marrow and continuous benzene exposure was also investigated in detail for simulated inhalation exposure concentrations ranging from 0.0039 ppm to 150 ppm. At the level of environmental exposures, no dose rate effect was found for either model. The occupational exposures led to only slight dose rate effects. A 32 ppm exposure for 15 minutes predicted consistently higher values than a 1 ppm exposure for eight hours for the total exposure of bone marrow to benzene and the cumulative quantity of metabolites produced by the bone marrow. The general relation between the cumulative quantity of metabolites produced by the bone marrow and the inhalation concentration of benzene is not linear. An inflection point exists in some cases leading to a slightly S shaped curve. At environmental levels (0.0039-10 ppm) the curve bends upward, and it saturates at high experimental exposures (greater than 100 ppm).

(Occup Environ Med 1994;51:414-420)

Benzene, a human carcinogen, is ubiquitous in the human environment. Workers in the petroleum, petrochemical, and other related industries are exposed through inhalation of benzene concentrations of the order of one part per million (ppm).¹⁻⁵ The general population is widely exposed to benzene in the atmosphere due to the volatile nature of petroleum, exhaust pipe emissions from motor vehicles, and emissions from other combustion sources. These exposures are typically in parts per billion (ppb).⁶⁷ Both workers and the general population are also exposed to benzene from cigarettes, either directly through smoking or indirectly from tobacco smoke in the environment. Within both populations, exposures can be nearly continuous or can involve intermittent peaks against some background level.

Estimates of the risks of benzene exposures in the general population are based on relatively simple extrapolations from high level exposures. Relations between exposure and tissue dose (and risk) can be non-linear, however, if saturable metabolic processes are involved. In such a case, to use the external exposure to the parent compound is incorrect. Physiologically based toxicokinetic models can provide more accurate estimates of risks by predicting tissue exposure to the active compound(s).

Many of the published experimental studies describe attempts to fit linear or compartmental toxicokinetic models to benzene toxicokinetic data. Recently, Travis et al developed a physiologically based toxicokinetic model for benzene in humans.8 No formal model parameter derivation was performed by Travis et al.8 Reference values were used for most of the model parameters. The metabolic parameters were then adjusted to visually fit the data. By contrast, in the work reported here, an attempt is made to formally fit a physiologically based toxicokinetic model. Multiple parameter sets were obtained that fit the data. These parameter sets were then used to investigate the toxicokinetics of benzene for realistic (environmental and occupational) exposures, focusing particularly on dose rate effects. One benefit of a physiological model is that it is reasonable to expect that extrapolations of such a model have more basis in reality than a purely empirical model. If the quantity of metabolites produced in the bone marrow is the relevant measure on which leukaemia risk estimates should be based, it is useful to have the relation between such a quantity and the exposure concentration. Thus target site exposures, difficult to measure experimentally, were also determined by simulation for a wide range of inhalation concentrations.

Methods

EXPERIMENTAL DATA

Data from four inhalation exposure studies were used to fit the model. Teisinger and Fiserova-Bergerova,⁹ as reported by Docter and Zielhuis,¹⁰ measured urinary phenol and its conjugates during and after exposure to 25 ppm of benzene for eight hours. The results

Department of Mechanical Engineering, University of California, Berkeley K H Watanabe D M Auslander

Lawrence Berkeley Laboratory F Y Bois J M Daisey

School of Public Health, University of California, Berkeley R C Spear

Requests for reprints to: Dr Frédéric Y Bois, Indoor Environment Program, mail stop 90-3058, Lawrence Berkeley Laboratory, Berkeley, California 94720, USA.

Accepted for publication 4 February 1994

are for a single subject, but this is not clearly indicated.

Srbova et al made measurements in 23 subjects, but reported the expired air and venous blood concentrations in only one person during a 90 minute exposure and up to 6.5hours after exposure.¹¹ The average benzene exposure concentration was about 98 ppm. The exact exposure concentrations reported at 15 minute intervals were used in the simulations, however.

Data reported by Sato et al for the average end tidal air and venous blood concentrations for three men were also used.12 The benzene exposure concentration was 25 ppm for a duration of two hours. The measurements after exposure were made for up to five hours. A second paper by Sato et al reported the mean and SD of benzene in end tidal air and venous blood concentrations after exposure to 25 ppm for two hours.¹³ The results were reported for five men and five women separately. The end tidal air measurements were made up to five hours after the exposure. Venous blood measurements were made both during and after exposure.

We code these studies as T for Teisinger and Fiserova-Bergerova,9 Sr for Srbova et al,11 Sa74 for Sato et al,12 and Sa75 for Sato et al.13 The results of Sato et al¹³ are treated as two separate experiments (men and women) for a total of five experiments used in the model parameter derivation.

MODEL STRUCTURE AND PARAMETER DERIVATION

A physiologically based toxicokinetic model, previously validated,14 was fitted to benzene concentrations in expired air and venous blood, and urinary phenol data. The model includes three compartments: central, bone marrow, and fat tissue compartment. Inhalation or ingestion exposures can be simulated. Benzene is eliminated by exhalation or through saturable (Michaelis-Menten) metabolism in the central and bone marrow compartments. The model requires the definition of 16 physiological variables calculated from their associated scaling coefficients. Table 1 gives the sampling ranges of the scaling coefficients, hereafter referred to as parameters. For fitting, the Monte Carlo technique of Spear et al¹⁴⁻¹⁶ was used. In a given Monte Carlo simulation, the parameters were selected randomly from uniform or log uniform distributions. The system of equations describing the model was solved numerically to make predictions for each criterion and the quality of the fit was assessed. The fit was declared acceptable if all predicted values (concentrations or metabolite amounts) were within 50% of the corresponding value of published data. Three to four experimental data points were chosen to summarise the time course of measurements made during exposure and after exposure. A table of the criteria is available on request. The parameter sets yielding good fits (meeting these criteria) were saved and labelled PASS parameter sets.

To improve sampling efficiency (finding PASSs), 3000 preliminary simulations were performed. From these 3000 runs, PASS parameter sets providing good fits to the data for each of the five experiments, considered individually, were extracted. The bounds of the parameter ranges (the lowest and the highest PASS parameter values) were determined. The intersection of the PASS parameter ranges for the five experiments was used for the parameter sampling in the subsequent Monte Carlo runs. Simulations were performed until 20 PASS parameter sets were found.

A cluster analysis was performed on the final PASS parameter sets obtained through Monte Carlo simulations to verify whether or

Table 1 Physiological variables and their scaling coefficients (SCs) for benzene toxicokinetics in humans, with the corresponding initial Monte Carlo sampling ranges*+

Physiological variable	SCs	Multiplier	SC lower limit	SC upper limit
Cardiac output	Sc-Flow-tot	BW ^{0.75}	0.188	0.543
Alveolar ventilation	VPerf-rat	Cardiac output	0.500	1.50
Blood flows:				
Bone marrow	Flow-bm	Cardiac output	0.00821	0.0665
Fat	Flow-fat	Cardiac output	0.0203	0.0784
Central compartment	Flow-cen	Cardiac output	±	‡
Volumes:		•	•	•
Bone marrow	V-bm	BW	0.010	0.060
Fat	V-fat	BW	0.109	0.337
Central compartment	V-cen	BW	S	S
Blood/air partition coefficient	PCb-art	1	1.66	Ĩ7·9
Tissue/blood partition coefficients:				
Bone marrow	PCb-bm	1	3.03	29.3
Fat	PCb-fat	1	23.0	70·5
Central compartment	PCb-cen	1	2.02	20.1
Maximum rate of metabolism, Vmax:				
Central compartment	Vab-cen	BW ⁰⁻⁷⁵	0.00104	0.170
Bone marrow	Vab-bm	Vab-cen	0.020	0.300
Vmax/Km ratios:				
Central compartment	Kab-cen	1	0.00533	0.362
Bone marrow	Kab-bm	1	0.000358	0.495
Elimination rate constant	Km-out	1	0.000211	0.0028
Fraction of metabolites excreted as phenol or phenol conjugates	Ph fraction	1	0.600	1.00

*Physiological parameter = scaling coefficient × multiplier. BW = body weight (kg), flows (l/min), volumes (l), Vmax (mg/min), Vmax/Km (l/min)

Initial lower and upper limits were obtained from scientific literature or scaled from animals. All scaling coefficients were sampled

from previous uniform distributions except where otherwise noted. ‡Values for this parameter were computed at each run so that the sum of the flows was equal to 100% of the total flow. §Values for this parameter were computed at each run so that the sum of the volumes was equal to 90% of the body volume. The variable was sampled with a log uniform distribution (uniformly sampled after log transformation).

Table 2 Exposures tested for predictions of the quality of metabolite formed in the bone marrow and the bone marrow exposure to benzene.

Exposures	Continuous background concentration (ppm)†	Peak exposure concentration (ppm)†	Peak exposure length (min)	TWA benzene exposure (mg/m³)
Environmental:				
I Environmental background‡	0.0039			0.0125
II Petrol pumping	0.0036	0.092	15	0.0125
III Cigarette smoking	0.00062	0.084	20	0.0132
Occupational:				
IV OSHA PEL		1.0	480	0.764
v		32.0	15	0.764

*Body weight of 70 kg was used for all simulations. †To convert to mg/l multiply by 0.003207, assuming $T = 25^{\circ}C$ and P = 1 atm. ‡Indoor average urban exposure.⁶

Peak exposures at time 0 and every 5040 min thereafter.

Two cigarettes smoked, at 0600, noon, and 1800 every day during the week.6

Table 3 Parameter range results obtained from the Monte Carlo sampling*

Scaling coefficient	T-Srt		Sa74-Sa75‡	
	Lower limit	Upper limit	Lower limit	Upper limit
Sc-flow-tot	0.297	0.477	0.38	0.52
VPerf-rat	0.547	1.08	0.725	1.17
Flow-bm	0.00969	0.0613	0.012	0.0612
Flow-fat	0.0204	0.0759	0.0531	0.0757
V-bm	0.0193	0.0592	0.0231	0.0587
V-fat	0.138	0.313	0.125	0.31
PCb-art	6.5	10.9	12	17.1
PCb-bm	6.31	27.1	5.14	27.3
PCb-fat	25.4	68	24.2	61.9
PCb-cen	2.02	2.62	2.01	2.63
Vab-cen	0.017	0.168	0.0316	0.167
Vab-bm	0.0452	0.271	0.0528	0.281
Kab-cen	0.00778	0.0169	0.0113	0.0209
Kab-bm§	0.000475	0.128	0.000505	0.0804
Km-out	0.0022	0.00268	0.000996	0.00263
Ph fraction	0.607	0.786	0.624	0.956

*Sampled uniformly except where otherwise noted. †T-Sr: Teisinger and Fiserova-Bergerova⁹ and Srbova *et al.*¹¹ ‡Sa74-Sa75: Sato *et al.*¹²¹³

Sampled log uniformly.

not they were connected in the parameter space, and thus formed one set of solutions.¹⁷

EXPOSURE SCENARIOS

After deriving parameters for the model, simulations were performed to predict the quantity of metabolites produced in bone marrow (Omet-bm) over one week at a steady state for various exposure scenarios. Predictions over the same week were also made for the integral of the concentration of benzene in bone marrow v time curve (area under the curve) which represents the total exposure of the bone marrow to benzene. Five realistic exposure scenarios were developed for the general population (ppb exposures) and for a worker population (ppm exposures). These were selected to represent a wide range of exposures, from the average population exposure to the occupational exposure at the current standards. For the general population, continuous and intermittent peak exposure scenarios that provided the same total inhalation dose were constructed (I and II, table 2). An air concentration of 0.0039 ppm benzene was assumed for the continuous background exposure scenario. This concentration is roughly the 80th percentile of the personal exposures measured for 50 people in Los Angeles in May, 1984.6 These 50 people were randomly selected to represent a population of 330 000 residents of the South Bay section of Los Angeles.

The second intermittent exposure scenario, the petrol pump scenario, assumes a continuous background exposure of 0.0036 ppm benzene in air. This models a subject refuelling his or her vehicle with unleaded petrol at 0700 on Monday and 1900 on Thursday. The refuelling process lasts 15 minutes and the average concentration of benzene in air is 0.092 ppm. This concentration was the geometric mean of the benzene concentrations measured in short term personal air samples of service station attendants in three locations in the United States.7

The third scenario, encountered by some members of the general population, was based on benzene exposures of a light smoker (six cigarettes a day) representing a total inhalation target dose of 300 μ g a day. Main stream cigarette smoke provides a benzene dose of about 50 μ g per cigarette based on measurements of main stream emissions of 1R4F cigarettes, a reference cigarette.^{18 19} This exposure scenario assumes a continuous background concentration of 0.00062 ppm benzene in air, which is at about the 10th percentile of the personal exposures measured for Los Angeles residents in May, 1984.6 Against that background, the smoker was assumed to smoke six cigarettes a day, two at 0600, at noon, and at 1800, 10 minutes exposure a cigarette, 50 μ g benzene a cigarette.

For the worker population, a continuous exposure of 1 ppm over an eight hour work day and a peak exposure of 32 ppm for 15 minutes a day were selected (IV and V table 2). These model exposures give the same total

Figure 1 Quality of fit. Model predicted v observed (experimental) data values. The bounds of 50% are indicated by the thin lines. (A) Data from Teisinger and Fiserova-Bergerova⁹ and Srbova, et al.11 Twenty predictions were made for each of 10 observed values. (B) Data from Sato et al^{12 13} Twenty predictions were made for each of 26 observed values.







exposure (480 ppm \times min). The 1 ppm exposure is the current Occupational Safety and Health Administration (OSHA) permissible exposure level (PEL). The 32 ppm for 15 minutes exposure would be in violation of the OSHA short term exposure level (STEL) of 5 ppm for 15 minutes.

CUMULATIVE QUANTITY OF METABOLITES V INHALATION CONCENTRATION OF BENZENE After parameter derivations of data for exposures of 25 and 98 ppm benzene, our model was used to make predictions of Qmet-bm at lower and higher exposure concentrations. These concentrations ranged from 0.0039 to 150 ppm and were simulated as continuous exposures.

Results

DERIVATION OF PARAMETER SETS

Parameter sets producing simulation results that pass all the goodness of fit criteria were not found. Two groups of parameter sets were obtained: one that fits both T and Sr data; and another group that fits Sa74 and Sa75 data. When the T-Sr PASS parameter sets are used to simulate the Sa74-Sa75 data, four predicted values are always below the 50% criteria limits. Similarly, when the Sa74-Sa75 PASS parameter sets are used to simulate the T-Sr data, three predicted values are always too high. The cluster analysis confirmed this result. Nearly 2.7 million simulations were needed to find 20 parameter sets fitting the T-

Figure 3 Quantity of metabolite formed in bone marrow over one week at steady state v the inhalation concentration. First five out of 20 curves obtained by fitting Sa74 and Sa75 data.



Sr data. Fitting Sa74 and Sa75 was easier. About 500 000 runs were needed to obtain 20 parameter sets that fitted.

Figure 1(A) shows the 20 predicted v observed values for all T-Sr data, with the model fits to these data (the data represent different measurements expressed in different units). A perfect fit to the data would have all the points on the diagonal (y = x). Figure 1(B) shows the 20 predictions for all Sa74 and Sa75 data, with the model fits to these data.

EXPOSURE SIMULATIONS

Figure 2 (A and B) plots Qmet-bm for exposure scenarios I-V. A set of connected markers represents the output from the model from one PASS parameter set. Exposure scenarios I and II have the same exposure time weighted average (TWA) and identical results predicted for Qmet-bm; thus no dose rate effects occur. The Qmet-bm for exposure V is consistently higher than the Qmet-bm values for exposure IV. The relative differences range from 0.13% to 4.6% (predictions from both T-Sr and Sa74-Sa75 derivations). The predictions for areas under the curves followed similar trends. No difference was seen between the areas under the curves for exposures I and II. The relative differences for exposures IV and V ranged from 0.20% to 11% based on predictions from both parameter derivations, therefore showing slight exposure rate effects.

CUMULATIVE QUANTITY OF METABOLITES V

INHALATION CONCENTRATION OF BENZENE Figure 3 is a plot of Qmet-bm v exposure concentration generated by the first five parameter sets fitting Sa74-Sa75 data. Of the 20 parameter sets, 11 sets generate S shaped curves such as runs 1 and 2. Four curves look like run 3, which increases more than linearly with concentration. The rest look like curves 4 and 5 where no S shape is seen and saturation is beginning. Of the 20 curves generated by T-Sr parameter sets, eight curves look like runs 1 and 2, two curves look like run 3, and seven curves behave like runs 4 and 5. In this group there were three parameter sets that generate seemingly straight lines over the range of concentrations studied. Figure 4 plots the slopes calculated between adjacent points v the mean inhalation concentration between the two points. From this plot it is Figure 4 Slope between each pair of adjacent points in fig 3 v the mean concentration between these two points.



clear that at low concentrations (< 50 ppm) the quantity of metabolite produced increases disproportionately with exposure concentration (increasing slopes seen in fig 4). Run 2, for example, predicts Qmet-bm equal to 0.114 mg/week at 0.005 ppm. At a concentration of 50 ppm Qmet-bm is 1494 mg/week. Above 50 ppm, the slope decreases slightly in most cases. Run 2 is an extreme case where the saturation effect at high dose is clear and the slope goes to zero above 150 ppm.

We attempted analytically to derive the functional relation between the rate of metabolite production by the bone marrow and the input concentration at steady state. Yet, due to the complexity of the system we were unable to obtain an explicit solution (even with the help of symbolic manipulation programs such as Mathematica and Maxima). A numerical solution was, however, obtained that verified the accuracy of the numbers computed by integration to within 0.33%.

Discussion

Experimental studies usually involve high exposure concentrations relative to what is encountered in the environment.⁹ ¹¹⁻¹³ ²⁰⁻²⁷ This is necessary to detect benzene concentrations in the blood and exhaled air of the subjects when assay sensitivity is low. The quantities typically measured in human studies are the benzene concentrations in venous blood and expired air, and the concentration of phenol in urine. The four studies used in this investigation were selected from the body of human data because we judged their reports to be the most complete in terms of exposure procedure and reliable analytical methods.⁹ ¹¹⁻¹³

Occupational studies are more realistic than experimental studies, but uncontrollable factors (for example, fluctuating and uncertain exposure concentration and differences in activity levels) make these data more uncertain.¹⁻⁵ The measured quantities are usually the same as in an experimental study, but with fewer and unevenly spread data points over time. For these reasons, analysis of these data, and in particular modelling them for prediction purposes, is difficult and may give results with greater uncertainty and variability than modelling experimental studies. We therefore did not use data from occupational studies.

MODEL STRUCTURE AND PARAMETER DERIVATIONS

Two distinct sets of data led to two different parameter derivations of the model. Therefore the two groups of data: group 1 consisting of T and Sr; and group 2 consisting of Sa74 and Sa75 seem to be incompatible. In terms of model parameters this translates into significantly different values for some of them. Table 3 shows that the univariate ranges obtained do not overlap for the blood to air partition coefficient. Also, it is likely that high dimensional correlation between parameters exists, distinguishing one PASS region from the other.

It is interesting that for both sets of parameters, the central compartment tissue to blood partition coefficient, PCb-cen, must fall in the range of 2.00 to 2.63 for both groups. This range covers only 7% of the sampled range. A true test of the model would be experimental verification of this result. Homogenised rabbit tissue to blood partition coefficients, for some of the tissues included in the central compartment, range from 1.08to 1.93,¹² a narrow range, but specific human values are unavailable.

The cause of the disparity between the two groups of model parameter sets is unclear. Some possibilities include: (a) the data simply represent two different underlying populations; (b) some measurements are poor or biased; (c) unreported differences in the experimental procedures exist; (d) the model structure is inadequate to describe the system dynamics; and (e) the limits for the criteria do not adequately account for the measurement uncertainty and between individual variability.

If the model structure is correct, then some conclusions about the data can be reached. Firstly, the differences in the blood to air partition coefficient could be indicative of pharmacogenetic differences in the subjects. Sato et al conducted their experiments in Japan,¹²¹³ whereas the T and Sr experiments were performed in Czechoslovakia.911 Pharmacogenetic differences in the subjects could be responsible for the way in which benzene was absorbed, metabolised, eliminated. or Secondly, changes in the analytical measurement technology over the 20 year period, or different techniques used by the the researchers could also account for the disparate results. We have carefully reviewed the measurement and experimental procedures, however, and see no obvious differences that would account for the two parameter derivations.

There is always the possibility that the model structure is inadequate and validation of this model as a reliable predictor can only take place when additional data become available. For now, one can only postulate the meaning(s) of the results presented here. In previous work the toxicokinetic models were considered valid if they were able to visually fit the data.^{8 28-31} These models were used for risk assessment^{30 31} although they were not validated with data pertaining to the quantity of interest. In this context the model presented here is of comparable quality.

The implication for risk assessment is that one should compare systematically the fits and model predictions obtained with different studies. This implies that the models should be formally fitted, with statistical techniques—for example, Monte Carlo simulations. Statistical fitting also gives confidence limits around model predictions, which should be a standard output of risk assessment.

CRITERIA AND THEIR MEANING

Confidence intervals, or more generally the distribution of predictions, represent several nested levels of variability: at least, analytical measurement errors, and within and between subject variability. It would be useful to have an estimate of population variability to judge whether results obtained in a Japanese study can be used for a Czechoslovakian population, for example. Yet, physiologically based models previously developed for the purpose of predicting doses to target tissues do not try to account for population variability. We propose that if the criteria for goodness of fit, as we define and use them here, account for measurement uncertainty and population variability, then both these effects are captured by the set of model parameters satisfying the criteria. As a consequence, each parameter derivation can be thought of as representing a possible member of the human population. The Monte Carlo method used here can therefore be presented as a "population toxicokinetic" approach.

The criteria were chosen to allow for population variability and uncertainty in the data. Sato *et al* reported SDs for their data.¹³ As this information was not available for all of the data, a standard within 50% was used. This is an underestimate of the variability in the urinary phenol data as Teisinger and Fiserova-Bergerova reported fivefold differences in values between subjects.⁹ For the end tidal air and blood benzene concentrations reported by Sato *et al*¹³ 50% more or less of the measured value is generally greater than two SDs of the mean for five subjects.

To assess the robustness of our findings, in particular the existence of two groups of data sets, with respect to the definition of the criteria, we attempted to fit the model across all data sets (T, Sr, Sa74, and Sa75) while removing three points that were less than a factor of two above the reported analytical detection limits. No PASS point for these new criteria was found by Monte Carlo sampling (1 250 000 simulations) and the clustering analysis again found two clusters.

SIMULATION RESULTS

Exposure rates are important to consider in regulatory policy because high exposure concentrations for short periods of time could result in different toxic effects than low exposures over long periods of time. Different rates of exposure were simulated to see if there was any difference in the predicted value for Qmet or area under the curve of benzene in bone marrow. As saturable reaction kinetics are known to describe the metabolism of benzene in the bone marrow and central compartments, it is possible that high concentrations for short periods of time could saturate the metabolic process, thereby producing dose rate effects.

For environmental exposures no dose rate effect was seen in the predictions of primary metabolites or for the benzene area under the curve in bone marrow. A slight but consistent dose rate effect appears at the level of occupational exposures (scenarios IV and V). The possibility of not having reached a steady state for exposure scenario IV was investigated by performing simulations up to five weeks. (As this scenario predicts the smaller values, not having reached steady state means that benzene is still accumulating in the bone marrow. The predicted values would only increase for exposure V if it was not at a steady state after three weeks.) The predicted values between the fourth and fifth week were the same as those reported between the second and third weeks.

The model fitted to the Sa74-Sa75 data gives predictions of Qmet-bm with less variability than the model fitted to the T-Sr data. The distribution of these predictions seems close to log-uniform. On the other hand, more variability was found in the prediction of area under the curve in bone marrow from the Sa74-Sa75 parameter sets. The predictions from both parameter sets seem to be uniformly distributed for areas under the curve for bone marrow. Although two distinct sets of model parameters were found, the predictions from both sets at low exposure concentration are similar (fig 2).

The predictions of Omet made by both derivations from the model are different from the findings of Bois and Paxman for rats exposed to benzene.³² In rats, no dose rate effects were found at the primary metabolite level that we studied here. At the level of individual metabolites, however, important dose rate effects were found. Slight dose rate effects are already present in the disposition of primary metabolites in humans. These may not be biologically significant by themselves, but it can be suspected that individual active secondary metabolites are even more affected, with possibly a large impact on cancer risk. In this context, the application of a short-term exposure limit for benzene seems warranted, even if it is difficult to assess whether its current value (5 ppm for 15 minutes) is over or under protective. More data on human metabolism of benzene is needed to precisely answer this question.

For continuous exposures, Qmet-bm predictions by Sa74-Sa75 parameter sets are generally greater than the predictions made by T-Sr parameter sets. Of the 20 Sa74-Sa75 predicted curves, nine of them showed generation of metabolites in excess of 1500 mg a week within the exposure concentrations studied. Only one out of 20 T-Sr curves showed production of metabolites in excess of 1500 mg a week.

Variability is not only present between the two groups of parameter sets but also within each group. Thinking of each parameter set as a possible member of the population, we see that some subjects exhibit saturation of the production of metabolites in bone marrow, whereas others increase linearly or superlinearly. In the group represented by the Sa74-Sa75 parameter sets an exposure of 25 ppm produces a ninefold difference between the highest and lowest Qmet-bm. There is a 24-fold difference in the T-Sr results at the same exposure concentration.

In recent risk assessments, it has been assumed that simple Michaelis-Menten kinetics describe correctly the relation between the applied dose and the mg equivalents of benzene metabolised per unit time.33 34 This study shows that Michaelis-Menten kinetics do not describe the relation at the site of action, the bone marrow. How, then, do we best extrapolate the high dose data to low doses in this context? The best approach should be the use of a general model, similar to ours, which can characterise the range of individual responses. A Michaelis-Menten approximation is easy to fit and would be appropriate for some individuals, but not for all. There is no guarantee that a simple Michaelis-Menten relation is even correct for extrapolating the population average. Assessment of cancer risk with average data should actually be avoided.

The findings described here: dose rate effects at occupational dose levels, S shape form of the production of metabolites as a function of exposure concentrations, large between individual variability, by their potential importance for human risk assessment of benzene induced leukaemia, deserve further attention. The recent publication of a new set of data²⁷ offers the potential for an independent verification of these results. Also, attempts are being made to acquire the missing information that made some of the older published reports unsuitable for this modelling investigation.

This work was supported by NIEHS grant number P42 ES04705-06 and by the Director, Office of Energy Research, Office of Health and Environmental Research, Human Health and Assessments Division of the United States Department of Energy under Contract No DE-AC03-76SF00098

- Sherwood RJ. Benzene: the interpretation of monitoring results. Ann Occup Hyg 1972;15:409-21.
 Inoue O, Seiji K, Kasahara M, Nakatsuka H, Watanabe T, Yin S-G, et al. Quantitative relation of urinary phenol levels to breathzone benzene concentrations: a factory
- ieveis to breamzone benzene concentrations: a factory survey. Br J Ind Med 1986;43:692-7.
 Drummond L, Luck R, Afacan AS, Wilson HK. Biological monitoring of workers exposed to benzene in the coke oven industry. Br J Ind Med 1988;45:256-61.
 Perbellini L, Faccini GB, Pasini F, Cazzoli F, Pistoia S, Rosellini R, et al. Environmental and occupational expo-sure to benzene by angling of breath and blood. Be V Ind.
- Sure to benzene by analysis of breath and blood. Br J Ind Med 1988;45:345-52.
- Med 1988;45:345-52.
 Brugnone F, Perbellini L, Faccini GB, Pasini F, Danzi B, Maranelli G, et al. Benzene in the blood and breath of normal people and occupationally exposed workers. Am J Ind Med 1989;16:385-99.

- 6 Wallace LA, Pellizzari ED, Hartwell TD, Davis V, Michael LC, Whitmore RW. The influence of personal activities on exposure to volatile organic compounds. Environ Res 1989;50:37-55. 7 McDermott HJ, Vos GA. Service station attendants' expo-
- sure to benzene and gasoline vapors. Am Ind Hyg Assoc J 1979:40:315-21.
- Travis CC, Quillen JL, Arms AD. Pharmacokinetics of benzene. *Toxicol Appl Pharmacol* 1990;102:400-20.
 Teisinger J, Fiserova-Bergerova V. Valeur comparée de la détermination des sulfates et du phénol contenus dans l'urine pour l'évaluation de la concentration du benzène
- Furine pour l'evaluation de la concentration du benzene dans l'air. Archives des Maladies Professionnelles et de Medicine du travail 1955;16:221-32.
 10 Docter HJ, Zielhuis RL. Phenol excretion as a measure of benzene exposure. Ann Occup Hyg 1967;10:317-26.
 11 Srbova J, Teisinger J, Skramovsky S. Absorption and elim-ination of inhaled benzene in man. Archives of Industrial Hygene and Occupational Medicine 1950;2:1-8.
 12 Sato A, Nakajima T, Fujiwara Y, Hirosawa K. Pharmacokinetics of benzene and toluene. Int Arch Arbeitsmed 1974;33:169-82.
- Arbeitsmed 1974;33:169-82.
- Arbeitsmed 19/4;33:109-82.
 Sato A, Nakajima T, Fujiwara Y, Murayama N. Kinetic studies on sex difference in susceptibility to chronic benzene intoxication—with special reference to body fat content. Br § Ind Med 1975;32:321-8.
- 14 Woodruff TJ, Bois FY, Auslander D, Spear RC. Structure and parameterization of pharmacokinetic models: their impact 189–201 on model predictions. Risk Anal 1992;12:
- 15 Bois FY, Woodruff TJ, Spear RC. Comparison of three physiologically based pharmacokinetic models of ben-tion of the pharmacokinetic models of the pha ene disposition. Toxicol Appl Pharmacol 1991;110: 79-88
- 16 Spear RC, Bois FY, Woodruff T, Auslander D, Parker J, Selvin S. Modeling benzene pharmacokinetics across three sets of animal data: parametric sensitivity and risk implications. *Risk Anal* 1991;11:641-54.
 17 Li H, Watanabe K, Auslander D, Spear RC. Model para-meter actination. undertanding accompting structure
- meter estimation: understanding parametric structure.
- meter estimation: understanding parametric structure. Ann Biomed Eng 1994(submitted).
 18 Brunnemann KD, Kagan MR, Cox JE, Hoffmann D. Determination of benzene, toluene and 1,3-butadiene in cigarette smoke by GC-MSD. Experimental Pathology 1989;37:108-13.
 19 Byrd GD, Fowler KW, Hicks RD, Lovette ME, Borgerding MF. Isotope dilution gas chromatography-mass spectrometry in the determination of benzene, toluene, stypene and activalization in mainstream ciacasta
- mass spectrometry in the determination of benzene, toluene, styrene and acrylonitrile in mainstream cigarette smoke. *J Chromatogr Sci* 1990;503:359-68.
 Hunter CG, Blair D. Benzene: pharmacokinetic studies in man. Ann Occup Hyg 1972;15:193-9.
 Nomiyama K, Nomiyama H. Respiratory retention, uptake and excretion of organic solvents in man. Int Arch Arbeitsmed 1974;32:75-83.
 Nomiyama K, Nomiyama H. Respiratory elimination of organic solvents in man. Int Arch Arbeitsmed 1974;32:

- organic solvents in man. Int Arch Arbeitsmed 1974;32: 85-91
- 23 Sherwood RJ. Comparative methods of biologic monitoring of benzene exposures. In: Proceedings of The Annual Conference on Environmental Toxicology. Fairborn, Ohio: National Technical Information Service, 1972:29-52.
- 24 Sherwood RJ. Pharmacokinetics of benzene in a human after exposure at about the permissible limit. Ann NY Acad Sci 1988;534:634-47.
- 25 Berlin M, Holm S, Knutsson P, Tunek A. Biological threshold limits for benzene based on pharmacokinetics of inhaled benzene in man. Arch Toxicol 1979; suppl 2: 305-310.
- 26 Berlin M, Gage J, Gullberg B, Holm S, Knutsson P, Tunek A. Breath concentration as an index of the health risk from benzene. Scand J Work Environ Health 1980;6:104-11.
- Pekari K, Vainiotalo S, Heikkilä P, Palotie A, Luotamo M, Riihimäki V. Biological monitoring of occupational exposure to low levels of benzene. Scand J Work Environ Health 1992;18:317–22.
- Ramsey JC, Andersen M. A physiologically based description of the inhalation pharmacokinetics of stryene in rats and humans. *Toxicol Appl Pharmacol* 1984;73:159–75.
 Paustenbach DJ, Clewell III HJ, Gargas ML, Andersen ME. A physiologically based pharmacokinetic model for inblade orthoact tetrachleride. *Toxicol Actor Discoursed* 1984;73:159–75.
 - inhaled carbon tetrachloride. Toxicol Appl Pharmacol 1988;**96**:191-211
- Reitz RH, McCroskey PS, Park CN, Andersen ME, Gargas ML. Development of a physiologically based pharmacokinetic model for risk assessment with 1,4-
- pharmacokinetic model for risk assessment with 1,4-dioxane. Toxicol Appl Pharmacol 1990;105:37-54.
 11 Leung H, Paustenbach DJ. Cancer risk assessment for dioxane based upon a physiologically-based pharmacoki-netic approach. Toxicol Lett 1990;51:147-62.
 22 Bois FY, Paxman DG. An analysis of exposure rate effects for benzene using a physiologically based pharmacoki-netic model. Regul Toxicol Pharmacol 1992;15:122-36.
 23 Bailer AJ, Hoel DG. Metabolite-based internal doses used in a risk assessment of benzene. Environ Health Perspect
- in a risk assessment of benzene. Environ Health Perspect
- a risk assessment of benzene. Environ Team Terspect 1989;82:177-84.
 Beliles RP, Totman LC. Pharmacokinetically based risk assessment of workplace exposure to benzene. Regul Toxicol Pharmacol 1989;9:186-95.