Estimating the Time of Last Cannabis Use from Plasma $\Delta^9$-Tetrahydrocannabinol and 11-nor-9-Carboxy-$\Delta^9$-Tetrahydrocannabinol Concentrations

Marilyn A. Huestis,$^1*$ Allan Barnes,$^1$ and Michael L. Smith$^2$

**Background:** Knowing the time cannabis was last used is important for determining impairment in accident investigations and clinical evaluations. Two models for predicting time of last cannabis use from single plasma cannabinoid concentrations—model I, using $\Delta^9$-tetrahydrocannabinol (THC), and model II, using the concentration ratio of 11-nor-9-carboxy-THC (THCCOOH) to THC—were developed and validated from controlled drug administration studies. Objectives of the current study were to extend the validation by use of a large number of plasma samples collected after administration of single and multiple doses of THC and to examine the effectiveness of the models at low plasma cannabinoid concentrations.

**Methods:** Thirty-eight cannabis users each smoked a 2.64% THC cigarette in the morning, and 30 also smoked a second cigarette in the afternoon. Blood samples (n = 717) were collected at intervals after smoking, and plasma THC and THCCOOH concentrations measured by gas chromatography–mass spectrometry. Predicted times of cannabis smoking, based on each model, were compared with actual smoking times.

**Results:** The most accurate approach applied a combination of models I and II. For all 717 plasma samples, 99% of predicted times of last use were within the 95% confidence interval, 0.9% were overestimated, and none were underestimated. For 289 plasma samples collected after multiple doses, 97% were correct with no underestimates. All time estimates were correct for 77 plasma samples with THC concentrations between 0.5 and 2 $\mu$g/L, a concentration range not previously examined.

**Conclusions:** This study extends the validation of the predictive models of time of last cannabis use to include multiple exposures and low THC concentrations. The models provide an objective and validated method for assessing the contribution of cannabis to accidents or clinical symptoms.

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Cannabis is one of the most widely abused drugs in the world, second only to alcohol (1). As one might expect with a high prevalence of use in the general population, cannabis is often involved in accidents and other mishaps of operations that require skill. For example, Soderstrom et al. (2) reported in 1988 that 34.7% of 1023 patients admitted to a large city trauma center as a result of vehicular and nonvehicular accidents had cannabinoids present in their blood. In a 1992 national safety study, Crouch et al. (3) found that 13% of truck driver killed in traffic accidents in the United States had $\Delta^9$-tetrahydrocannabinol (THC), the primary active component in cannabis, or its metabolite 11-nor-9-carboxy-THC (THCCOOH) in their blood or urine at autopsy. These and similar studies have demonstrated a connection between accidents and cannabis use, but have not established causation by cannabis, despite the knowledge that individuals who use cannabis have impaired performance in driving simulator and on-the-road tests (4–8). Drummer et al. (11) did determine causality in a study of 3398 fatally injured drivers in Australia,
using culpability analysis, and reported that drivers with blood THC concentrations of 5 µg/L or greater were 6.6 times more likely than drug-free drivers to have a fatal accident. Medical and forensic investigators tasked with establishing causality and impairment use information from these and similar studies. Knowing the elapsed time from last drug use is necessary for applying these data to an individual case.

Scientists currently evaluate impairment in individual cases by estimating the time of drug use and relating these parameters to those found in pharmacodynamic studies (12). Unlike cases involving use of alcohol, impairment of performance from cannabis use is not easily correlated to blood or plasma concentrations of THC or its metabolites. Maximum drug effects may occur later than peak blood concentrations of THC or its active metabolites. Effects on the brain continue as blood concentrations of active drug decrease, a process termed hysteresis (13).

Investigations dating back to those by Lemberger (14) have attempted to relate blood or plasma cannabinoid concentrations with the time after smoking or oral ingestion of cannabis (14–19). Using different methods of analysis, scientists produced expected time-vs-plasma concentration profiles but described large inter- and intra-individual variations in peak values, times to reach maximum concentrations, and areas under the curve. Most persons who were infrequent cannabis users had plasma THC concentrations <1 µg/L after 4 h (20). Peat (21) introduced results for frequent cannabis users showing mean (SD) plasma THC and THCCOOH concentrations of 0.86 (22) and 45.8 (13.1) µg/L, respectively, for persons who reported their last cannabis use more than 12 h before blood collection. In a well-controlled study of 6 cannabis users, Huestis et al. (19) collected blood samples from volunteers after they had smoked single cannabis cigarettes containing 0%, 1.75%, and 3.55% THC. Participants were housed in a secure medical unit with dosing delayed until urine concentrations of cannabinoids were <20 µg/L. Conditions that contribute to interindividual variation in plasma concentrations of THC and its metabolites were carefully controlled, e.g., cigarette potency, number of puffs, time between puffs, length of inhalation, and length of time that smoke was held in the lungs were standardized, but for safety reasons other variables, such as depth of inhalation, were not. First samples were collected at 1 min after smoking and at frequent intervals up to 168 h. From these data, the authors developed 2 models to predict time of last cannabis use within 95% confidence intervals (CIs). The first model computed the elapsed time between smoking cannabis and blood collection based on plasma THC concentration alone, whereas the second model used the plasma THCCOOH/THC concentration ratio. They applied the models to results from all published studies at the time that reported THC and/or THCCOOH concentrations, measured by either RIA or gas chromatography–mass spectrometry with either internal or external standardiza-

### Materials and Methods

#### Participants and Study Design

Descriptions of the participants and study design were reported previously (24). Thirty-eight males with a history of cannabis use provided informed consent to this National Institute on Drug Abuse Intramural Research Program Institutional Review Board–approved protocol. The study participants were compensated for their time and effort. All resided on the secure clinical research unit for at least 1 day before dosing and 2 days afterward. No drugs were administered until the urine cannabinoid concentrations were <20 µg/L. For safety, participants were medically evaluated during and for 1 week after drug administration.

At 0900 on the day of testing, participants received a single oral dose of placebo (n = 10) or up to 90 mg of rimonabant. Two hours later, they smoked a cannabis cigarette containing 2.64% THC by weight, estimated to contain ~20 mg of THC. The number of puffs and time between puffs were standardized. Blood for THC and THCCOOH assays was drawn from an indwelling venous catheter in the arm 120 and 5 min before cannabis smoking and at 2, 5, 10, 15, 20, 25, 30, 50, 70, 90, 110, and 235 min after the start of smoking. Heparinized plasma was stored at −20°C until analysis. Four hours after smoking the first cigarette, 30 of the men smoked a second
cigarette containing 2.64% THC. Blood samples were collected at the same time intervals as described for the first dose, except that the 235-min collection was omitted.

The previously published study designed to investigate the blockade of cannabis effects demonstrated that rimonabant had no effect on THC and THCCOOH pharmacokinetics (24). Therefore, the participants receiving rimonabant and placebo could be grouped together for the current study and the results generalized to cannabis users.

Plasma THC and THCCOOH concentrations were determined in plasma blood collected from 38 participants in the morning smoking session and 30 who also smoked in the afternoon. Occasional samples had THC concentrations >2.5 \( \mu \text{g/L} \), the limit of quantification (LOQ) for the method. In all, 717 plasma THC and 704 plasma THCCOOH concentrations were evaluated with the 2 predictive models.

**Analytical Method**

Plasma samples were analyzed for THC and THCCOOH by a previously published method (25). Briefly, trideuterated THC and THCCOOH internal standards were added to plasma to improve identification and quantification. Proteins were precipitated with acetonitrile. The extraction method used CleanScreen solid-phase extraction columns (United Chemical Technologies), with THC eluted with hexane–ethyl acetate–ammonia hydroxide (93:5:2 by volume) and THCCOOH eluted with hexane–ethyl acetate (70:30 by volume). Trifluoracetyl-THC and trifluoroacetyl-hexafluoroisopropyl-THCCOOH derivatives in the same reconstituted extract were injected into a Finnegan/MAT Model 4023 gas chromatograph–mass spectrometer operated in the negative-ion chemical ionization mode (70:30 by volume) and THCCOOH eluted with hexane–ethyl acetate–ammonia hydroxide (93:5:2 by volume). Trifluoracetyl-THC and trifluoroacetyl-hexafluoroisopropyl-THCCOOH derivatives in the same reconstituted extract were injected into a Finnegan/MAT Model 4023 gas chromatograph–mass spectrometer operated in the negative-ion chemical ionization mode with methane as reagent gas, helium as carrier gas, and a DB 1 capillary column (J & W Scientific). Cannabinoid concentrations were determined by use of ion \( m/z \) 410 for THC and \( m/z \) 422 for THCCOOH. The LOQs were 0.5 \( \mu \text{g/L} \) for THC and 2.5 \( \mu \text{g/L} \) for THCCOOH, with CVs across the analytical range of 4.1%–11% and 4.9%–12%, respectively.

**Predictive Models**

Huestis et al. (19) previously published models for estimating time of last cannabis use. Model I determined time estimates from plasma THC concentrations and model II from the plasma THCCOOH/THC concentration ratios. The formulas are reproduced below, with \( t \) representing the elapsed time in hours between the beginning of cannabis smoking and blood collection, and CI representing the 95% confidence interval for the estimate of \( t \). The subscripts 1 and 2 refer to models I and II, respectively, and brackets indicate the concentrations of THC or THC-COOH in \( \mu \text{g/L} \).

**Model I:**

\[
\log t = -0.698 \log [\text{THC}] + 0.687
\]

\[\log \text{Cl}_1 = \log t \pm 1.975\]

**Model II:**

\[
\log t = (0.576 \log [\text{THCCOOH}]/[\text{THC}]) - 0.176
\]

\[\log \text{Cl}_2 = \log t \pm 1.975\]

These equations were developed from cannabinoid concentrations in plasma samples collected for up to 168 h after controlled smoking of a 1.75% and a 3.55% THC cigarette by each of 6 cannabis users residing continuously in a secure clinical research unit (19). Drug administration was not initiated until their urine cannabinoid concentrations were <20 \( \mu \text{g/L} \). Table 1 is reproduced from that report (19) and displays the 95% CIs, in hours, for selected predicted times after smoking using models I and II. These results are presented to show that the models predict last use within an interval of time. The magnitude of the interval is smaller when the elapsed time between cannabis use and blood collection is shorter.

In the current study, model I was applied to each plasma sample with a valid THC measurement and model II to those with valid THC and THCCOOH concentrations. All analyte results at or above the LOQs of the method were selected. For each model, the CI for time of last use was determined. The actual times of smoking were examined to determine whether they were within (i.e., correct) or outside (i.e., incorrect) the CI. The following parameters were calculated for each model: (a) accuracy [(number of correct time estimates/total number of estimates) \( \times \) 100%]; (b) the mean magnitude of error [mean of absolute values of \( t_{\text{actual}} - t_{\text{calculated}} \) for all incorrect estimates]; and (c) the number of samples with actual times greater than the upper limit of the CI (i.e., the model estimated a shorter \( t \) than actual, termed an underestimate) or below the lower limit (i.e., the model estimated a longer \( t \) than actual, termed an overestimate) for each situation examined.

We applied the same evaluations to each sample, using the models in combination. With this application, esti-

<table>
<thead>
<tr>
<th>Predicted elapsed time, h</th>
<th>95% CI, h</th>
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<tbody>
<tr>
<td>0.5</td>
<td>0.2–1.1</td>
</tr>
<tr>
<td>1</td>
<td>0.5–2.2</td>
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<tr>
<td>2</td>
<td>0.9–4.4</td>
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<tr>
<td>4</td>
<td>1.8–8.8</td>
</tr>
<tr>
<td>6</td>
<td>2.7–13.2</td>
</tr>
<tr>
<td>8</td>
<td>3.6–17.6</td>
</tr>
</tbody>
</table>

*a Reproduced from Huestis et al. [J Anal Toxicol 1992;16:283–90 (19)] by permission of Preston Publications, A Division of Preston Industries, Inc.*
mates were considered correct if the true value of $t$ fell within a range defined by the lower limit of the CI of either model and the upper limits of the CIs, whichever was highest.

**Results**

The results for predicted elapsed times between the beginning of cannabis smoking and blood collection obtained with model I, model II, and a combination of the models are summarized in Table 2. The conditions for the single cigarette study were similar to those in the study used to produce the models, except that cigarettes contained 2.64% THC instead of either 1.75% or 3.55% THC, and length of inhalation and the time smoke was held in the lungs were not controlled. For model I, Table 2 reflects that, for 392 of 427 samples collected from 38 individuals after they began smoking the first cigarette, the observed time fell within the predicted range of elapsed time. Another way to view this result is that the model was correct for 91.8% of cases. For those cases that were overestimated, the average error of actual elapsed times was 14 min lower than the lowest value in the CI. There were no underestimates when model I was applied to data obtained after the first cigarette.

As can also be seen in Table 2, the models provided similar results after volunteers ($n = 30$) smoked a second cannabis cigarette in the afternoon. A difference noted after multiple cigarettes was that 3 of 290 samples had actual elapsed times longer than predicted by model I, i.e., for these 3 samples the time of last use was underestimated.

Model II correctly estimated the time of last smoked dose for 94.0% of samples after a single cigarette. Unlike model I, most of the errors (21 of 415 samples) were underestimates. Estimates were more accurate after multiple doses.

We evaluated whether the CIs for models I and II could be combined to produce superior results. Shown in the lower portion of Table 2 are the accuracy, overestimates, and underestimates in predicted elapsed times when we used models I and II in combination. If the actual elapsed time after the beginning of smoking fell within a range with a lower limit equal to the lowest predicted confidence limit for either model and an upper limit equal to the highest predicted confidence limit for either model, then results were considered correct. Accuracies were near 99%, and there were no instances of underestimated predictions.

One of the limitations mentioned in the introduction was that the models had not been challenged for plasma THC concentrations $<2 \mu g/L$. This is an important limitation because concentrations in this range are expected in the first 4 h after cannabis smoking, the timeframe in which many individuals are impaired and attempt to perform skilled operations (20). In this study, 77 plasma samples from 26 individuals had THC concentrations in the range of 0.5–2 $\mu g/L$. The results of application of the models to these samples are shown in Table 3. All 77 predictions were accurate with the combination of models I and II. Independently, the models were not as accurate for these concentrations as they were for samples with concentrations $\geq 2 \mu g/L$: 80.5% compared with 91.2% for model I, and 77.6% compared with 95.5% for model II.

**Discussion**

For most forensic and medical applications, the ultimate goal of determining the time of last cannabis use is to relate this time to effects of the drug on the body. Effects vary among individuals, but most people experience the physiologic effects of increased heart rate and conjunctival injection or bloodshot eyes (26, 27). These effects begin during smoking and may last up to several hours. Effects on the brain include euphoria and decrements in memory, the ability to maintain attention, estimating time, and

| Table 2. Predicted elapsed time between cannabis smoking and blood collection with plasma THC $\geq 0.5 \mu g/L$ and THCCOOH $\geq 2.5 \mu g/L$. |
|-------------------------------------------------|-------------------------------------------------|-------------------------------------------------|
| **Model I** | **Model II** | **Combination of models I and II** |
| **Single dose** | **Multiple doses** | **All cases** | **Single dose** | **Multiple doses** | **All cases** | **Single dose** | **Multiple doses** | **All cases** |
| n | Accuracy (samples within 95% CI), % | n | Mean (range) time, min | n | Mean (range) time, min | n | Mean (range) time, min |
| 427 | 91.8 | 35 | 14 (<1 to 41) | 0 | 0 | 21 | 38 (2–94) |
| 290 | 90.3 | 25 | 18 (<1 to 50) | 3 | 5 (2–11) | 0 | 0 |
| 717 | 91.2 | 60 | 16 (<1 to 50) | 3 | 5 (2–11) | 21 | 38 (2–94) |
| 415 | 94.0 | 4 | 3 (<1 to 6) | 21 | 38 (2–94) | 0 | 0 |
| 289 | 97.6 | 7 | 2 (<1 to 4) | 0 | 0 | 21 | 38 (2–94) |
| 704 | 95.5 | 11 | 2 (<1 to 6) | 21 | 38 (2–94) | 0 | 0 |
| **Combination of models I and II** | **Combination of models I and II** | **Combination of models I and II** |
| **Single dose** | **Multiple doses** | **All cases** | **Single dose** | **Multiple doses** | **All cases** |
| n | Accuracy (samples within 95% CI), % | n | Mean (range) time, min | n | Mean (range) time, min |
| 415 | 99.5 | 2 | 1 (<1 to 1) | 0 | 0 |
| 289 | 98.6 | 4 | 2 (<1 to 4) | 0 | 0 |
| 704 | 99.1 | 6 | 2 (<1 to 4) | 0 | 0 |

* Time interval defined by the lowest and highest 95% confidence limits of both models.
significant effects the following day
ported subjective and impaired behavioral effects for up
debated among scientists. Some investigators have re-
of time that acute effects on the brain continue has been
coordination on divided attention tasks (23). A consensus panel of experts that met to determine the current state of knowledge for effects of cannabis on
driving performance reported that “most behavioral and physiological effects return to baseline levels within 3–5 h
after drug use, although some investigators have demonstrated residual effects in specific behaviors such as complex divided attention tasks for up to 24 h” (23). Using the models and an individual’s plasma THC and THCCOOH concentrations, one may estimate the time the individual last used cannabis and from this time predict cognitive or performance decrements.

The results of this study demonstrate that model I accurately predicts the time of last cannabis use for >90% of cases and that incorrect predictions are overestimates after single doses. After multiple doses of cannabis, accuracy remains high, but 3 of 290 results were underestimated. Although that number is small, we pay special attention to these underestimates because in general they have a greater impact on actual cases than do overestimates. For example, if a judicial investigator or physician wants to know whether a person with THC in his or her blood was impaired at the time of an accident, it is important to know how long before the accident that person smoked cannabis. Underestimating this time increases the potential for falsely accusing an individual of being impaired at the time of the accident. For our 3 cases, the underestimates were not serious problems, however, because the magnitudes were small, ranging from 2 to 11 min. If we used the estimated times in these cases to predict the effects of cannabis on the individuals, the errors in time are small compared with the variability in specific effects and will not impact our decision.

Determining accuracy after a second cigarette is an important finding because previous studies did not address applying the models to multiple THC administrations. Bogusz, in a letter to the editor in the Journal of Analytical Toxicology (34), questioned the applicability of the predictive models when individuals ingested multiple doses of cannabis. Huestis and Cone addressed many, but not all, of his concerns in a response and suggested that the models be examined in a multiple-dose study. We now demonstrate that model I gives accurate predictions for >90% of plasma samples collected after multiple cannabis doses.

After single doses, model II was 94.0% accurate, but most errors (21 of 415 samples) were underestimates. However, all 21 were from samples collected >3 h after smoking. Huestis et al. (19) noted in the publication of the models that predictions at longer elapsed times were occasionally underestimated. For those times that were underestimated, the magnitude ranged from 2 to 94 min with a mean of 38 min.

The combined models gave time estimates that were 99.1% accurate with no underestimates. In addition, when overestimates occurred, they were less than 4 min. For the subset of samples with THC concentrations between 0.5 and 2.0 μg/L and THCCOOH concentrations ≥2.5 μg/L, the combined-model approach was correct in all instances. This method of analysis yields a longer estimated time interval, but greater certainty in prediction. One important outcome of greater certainty is that underestimates of time of use were eliminated. For forensic cases in which finders of fact wish to avoid underestimating time of use, the combined CI model offers greater accuracy and is the method of choice.

The cannabis cigarettes used in our study contained 2.64% THC. Individuals smoked 1 puff/min but chose their own depth of inhalation. This rate is greater than self-paced smoking. All had increases in heart rate and subjective drug effects. According to ElSohly et al. (35), in 1997 the mean THC concentration in seized cannabis was 4.2%. Of course, cannabis can have a higher THC content. The models were developed from plasma data collected after persons smoked a 1.75% and a 3.55% cigarette, the latter having more THC than used in the current study (19). Although peak concentrations were higher for the 3.55% cigarettes, the times of use fell within a range similar to that for the lower-dose cigarette. Other studies also have shown that cannabis users tend to titrate their dose of drug to maintain the high they feel (20). One might expect users who are free to choose their own smoking pattern to keep the dose of THC within an effective range even if the cigarette THC content is high. Their plasma THC and THCCOOH concentration profiles

### Table 3. Predicted elapsed time between cannabis smoking and blood collection with plasma THC between 0.5 and 2.0 μg/L and THCCOOH ≥2.5 μg/L.

<table>
<thead>
<tr>
<th></th>
<th>Model I</th>
<th>Model II</th>
<th>Models I and II*</th>
</tr>
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<tbody>
<tr>
<td>n</td>
<td>77</td>
<td>76</td>
<td>76</td>
</tr>
<tr>
<td>Accuracy (cases within 95% CI), %</td>
<td>80.5</td>
<td>77.6</td>
<td>100.0</td>
</tr>
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<td>Elapsed time overestimated</td>
<td></td>
<td></td>
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<tr>
<td>n</td>
<td>15</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Mean (range) time, min</td>
<td>31 (3–50)</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Elapsed time underestimated</td>
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<tr>
<td>n</td>
<td>0</td>
<td>17</td>
<td>0</td>
</tr>
<tr>
<td>Mean (range) time, min</td>
<td>0</td>
<td>37 (5–87)</td>
<td>0</td>
</tr>
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</table>

* Time interval defined by the lowest and highest 95% confidence limits of both models.
would be similar to those after smoking of a less potent cigarette. In addition, the models have been shown to be accurate in many legal proceedings in multiple countries over the last 13 years when the times of cannabis use were known.

A common problem in applying the models to actual situations is that many accident victims die and THC and THCCOOH concentrations are available only for postmortem blood. As mentioned in the introduction, the average antemortem blood-to-plasma ratio is $\sim$0.5. One can estimate a plasma concentration from a blood concentration by dividing by 0.5 and inserting the plasma concentration into the formulas to estimate time of last cannabis use. However, in fatal accident cases, the important variables are the concentrations of drugs and metabolites in antemortem plasma; estimating these from postmortem blood has not been well documented. Giroud et al. (36) studied the concentrations of THC and THCCOOH in blood and plasma from live patients and found mean ratios similar to that reported above, specifically 0.67 and 0.59, respectively. These authors also measured postmortem blood and postmortem “serum” (obtained from centrifuging postmortem whole blood) cannabinoid concentrations and found variable ratios with mean blood-to-serum ratios of 0.45 for THC and 0.37 for THCCOOH. The causes of this variability can be many, including changes in serum after death and postmortem redistribution. The study did have an unavoidable limitation of being unable to relate postmortem blood and antemortem plasma analyte concentrations. Errors in converting postmortem blood to antemortem plasma concentrations may have an impact on predictions using model I but would have less effect on predictions using model II because model II estimates are based on the ratio of THCCOOH to THC. These 2 cannabinoid molecules are similarly partitioned between plasma and blood cells, making the ratio less sensitive to partition variability.

In conclusion, the validation of 2 previously described models for predicting the time of last cannabis use from plasma THC and THCCOOH concentrations was extended to samples collected after multiple cannabis exposures and also those with low THC concentrations. The predictive models provide an objective and validated method for assessing the contribution of cannabis to accidents or clinical symptoms based on cannabinoid concentrations in a single plasma sample.

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