

## In this issue . . .

### Tracking the source of radioactive ruthenium



An air sampler stationed at the top of Puy de Dôme in central France.

In October 2017, significant atmospheric levels of the radioactive isotope ruthenium-106 ( $^{106}\text{Ru}$ ) were detected in many countries throughout Europe. The detected levels, although too low to endanger human or environmental health, suggested a sizeable release of  $^{106}\text{Ru}$  not associated with declared nuclear accidents. O. Masson, G. Steinhauser, et al. (pp. 16750–16759) compiled more than 1,100 atmospheric measurements and 200 deposition measurements of  $^{106}\text{Ru}$  levels from across Eurasia. The vertical distribution of  $^{106}\text{Ru}$  was inconsistent with an alleged disintegration of a radioisotope-carrying satellite. The highest  $^{106}\text{Ru}$  levels were detected over Romania. However, the detections occurred at multiple locations throughout the country simultaneously, suggesting that the  $^{106}\text{Ru}$  plume had widened to the width of Romania by that point, and therefore originated a considerable distance away. Models of air mass movements in the days preceding detection and  $^{106}\text{Ru}$  deposition patterns were consistent with a release from the Mayak nuclear complex in the Southern Urals of Russia. The  $^{103}\text{Ru}/^{106}\text{Ru}$  ratio, combined with the absence of other radioisotopes, suggests that the  $^{106}\text{Ru}$  came from reprocessing of approximately 2-year-old spent nuclear fuel. According to the authors, the release might have occurred in conjunction with production of a cerium-144 source for a neutrino experiment. — B.D.

### Deep learning and protein structure prediction

Proteins consist of amino acid sequences, folded into 3D structures that dictate function. Predicting protein structure is key to understanding biological processes, but existing methods are slow and computationally costly, particularly for proteins that lack abundant sequence homologs. Jinbo Xu (pp. 16856–16865) presents an approach that uses deep learning to accurately predict the interresidue distance distribution of a protein from approximately 60 sequence homologs. Using the approach, the author generated a 3D model of folding for the protein based solely on the predicted geometrical constraints. Requiring only 4 hours on a Linux computer running 20 central processing units, the method accurately folded 21 of 37 so-called CASP12 hard targets—particularly challenging structures established by the automated protein structure prediction community—and surpassed by a wide margin a popular approach that does not use deep learning. In addition, the author's approach correctly folded

2 membrane proteins whose structures have defied prediction. Although further tests are needed, the findings suggest that protein structures can be accurately predicted with deep learning on a personal computer, according to the author. — T.J.

### Neonicotinoids, honeydew, and insect mortality

Neonicotinoids are among the most widely used insecticides and can harm beneficial insects that feed on contaminated nectar and pollen. Miguel Calvo-Agudo et al. (pp. 16817–16822) examined whether beneficial insects could also be exposed to neonicotinoids through contaminated honeydew. Honeydew is a sugar-rich substance excreted by phloem-feeding insects and an important nutrient source for many beneficial insects, such as pollinators and natural enemies of insect pests. The authors collected honeydew produced by citrus mealybugs (*Planococcus citri*) that fed on citrus trees treated with either water or the neonicotinoids

