

Supplementary Information

Appendix A HBV Model

The HBV model is a commonly used rainfall-runoff model with routines for different components of the water cycle (e.g., snow routine, soil routine, groundwater routine for response and routing, etc.) [33, 74, 75]. The snow routine of the HBV model was used to capture (1) the partitioning of precipitation, $P(t)$ [$L\ T^{-1}$], between rainfall, $r(t)$ [$L\ T^{-1}$], and snowfall, and (2) the accumulation of snowpack and generation of snow melt, $s(t)$ [$L\ T^{-1}$]. Model parameters include (1) the threshold temperature, T_T [K], below which all precipitation is considered to fall as snow, (2) the snowfall correction factor, SF_{cf} [-], which accounts for snowfall undercatch due to wind turbulence, catchment vegetation, etc. [74, 76], (3) the degree-day factor, D_F [$L\ T^{-1}\ K^{-1}$], which is a proportionality constant for estimating snow melt, (4) the water holding capacity of the snowpack, W_H [-], and (5) the refreezing coefficient, F [-], which allows for refreezing of the melted water when air temperature, $T_{air}(t)$ [K], falls below T_T . Model parameters for snow accumulation were chosen to mimic weather conditions in the Mid-Atlantic U.S. (Table A1). To validate the HBV snow model results, we used data on daily accumulated snow depth at the nearby Dulles Airport (Figure A1), downloaded from the National Oceanic and Atmospheric Administration website

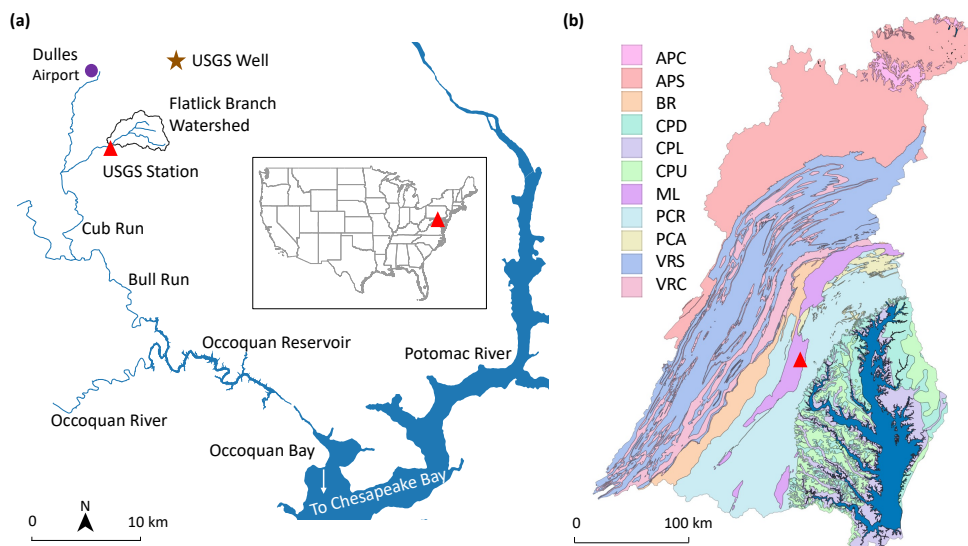


Fig. A1 (a) Flatlick Branch drains an urban watershed in Northern Virginia, and is tributary to the Occoquan Reservoir, where rising sodium concentrations threaten a critical water supply for up to 1 million people; (b) Flatlick Branch is located in the Mesozoic Lowland hydrogeomorphic (also referred to as the “Triassic Lowland”) region of the Chesapeake Bay watershed (APC/APS: Appalachian Plateau Carbonate/Siliciclastic, BR: Blue Ridge, CPD/CPL/CPU: Coastal Plain Dissected Uplands/Lowlands/Uplands, ML: Mesozoic Lowlands, PCR/PCA: Piedmont Crystalline/Carbonate, VRS/VRC: Valley and Ridge Siliciclastic/Carbonate).

(www.ncei.noaa.gov), to verify that the timing of predicted snow melt events aligned with the date of zero accumulated snow depth reported at Dulles Airport after snow events. From these comparisons we estimated that the HBV snow model accurately predicts snow melt at Dulles roughly 83% of the time.

Table A1 HBV model parameters.

Parameter	Value
T_T	0° C
SF_{cf}	1.1
D_F	$7.29 \times 10^{-5} \text{ m } ^\circ\text{C}^{-1} \text{ h}^{-1}$
W_H	0.1
F	0.05

Appendix B Hydrologic Model

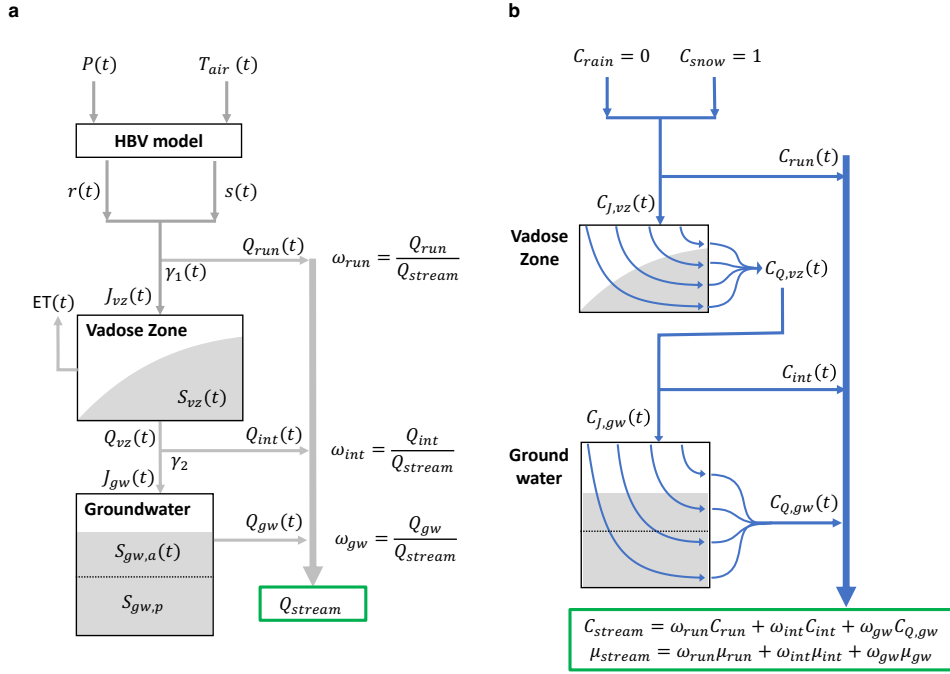


Fig. B2 Conceptual representation of the (a) hydrologic model and (b) transit time distribution model.

Stream flow was assumed to be the sum of direct runoff to the stream (e.g., through storm drain networks and overland flow), interflow through the vadose zone, and groundwater discharge: $Q_{\text{stream}}(t) = Q_{\text{run}}(t) + Q_{\text{int}}(t) + Q_{\text{gw}}(t)$ [L T^{-1}] (Figure B2a). The three terms on the right hand side of this equation were calculated as follows. A fraction, $\gamma_1(t)$ [-], of rain and snow melt (output from the HBV snow model) infiltrates to the vadose zone, $J_{\text{vz}}(t) = \gamma_1(t)(r(t) + s(t))$ [L T^{-1}], while the rest flows directly to the stream through storm drain networks and overland flow, $Q_{\text{run}}(t) = (1 - \gamma_1(t))(r(t) + s(t))$ [L T^{-1}] (Vadose Zone storage, Figure 1a). Preliminary evaluation of the model indicated that infiltration into the vadose zone varies seasonally, presumably due to seasonal variations in vegetation cover and rainfall intensity (discussed further below) [77]. Accordingly, the infiltration fraction, $\gamma_1(t)$, was allowed to vary non-linearly with day, d , of the water year: $\log_{10} \gamma_1(t) = x + i \sin \frac{2\pi(d+p)}{365}$. In this expression, $d = 0$ corresponds to the beginning of the water year, October 1st, and the parameters x [-], i [-], and p set, respectively, the annual average infiltration, magnitude of the seasonal signal, and phase of the seasonal signal. For the top model (with the lowest RMSE, see Table 1), the set of values inferred for x , i , and p suggest that the fraction of inflow that enters the vadose zone through infiltration is higher in the winter and lower in the summer.

We hypothesized that most of this seasonality arises from seasonal variations in the intensity of storm events; specifically, if summer storms are frequently more intense than winter storms, a larger fraction of summer-time rainfall would enter the stream as direct runoff, and a smaller fraction would enter the vadose zone through infiltration. An analysis of output from the HBV snow model (i.e., hourly estimates of inflow to the catchment from either rainfall or snow melt) supports this hypothesis. From the HBV snow model, we obtained ten years of hourly estimates for inflow to the catchment from either rainfall or snow melt ($N=87672$). We started our analysis by removing from this timeseries all instances where the HBV snow model predicted zero inflow (i.e., we removed all dry weather periods). From the remaining non-zero inflow intensity values ($N=7250$) we constructed three probability distributions, one for inflow during the summer (June, July and August), one for inflow during the winter (December, January, February), and one for inflow year round. Consistent with our hypothesis, the distributions for winter and summer inflow intensities peak around 0.2 and 0.5 mm h^{-1} , respectively, and very intense storms (up to 6 mm h^{-1}) are more frequent in the summer (Figure B3). The seasonal difference in inflow intensities can also be documented by examining the raw data. Over the ten year time period, the number of hours when summer and winter inflow intensities exceeded 2 mm h^{-1} were $N=320$ and 253, respectively. Conversely, the number of hours when summer and winter inflow intensities were below 2 mm h^{-1} were $N=685$ and 2479, respectively. Thus, lower intensity inflow events are more frequent in the winter than the summer, while higher intensity inflow events are more frequent in the summer than the winter. This analysis supports our hypothesis that the most of the seasonal variation in $\gamma_1(t)$ reflects runoff generation during intense summer rains.

Once water enters the vadose zone, $J_{\text{vz}}(t)$ [L T^{-1}], it can accumulate with time, $\frac{dS_{\text{vz}}}{dt}$ [L T^{-1}], leave by evapotranspiration, $ET(t)$ [L T^{-1}], or leave by discharge, $Q_{\text{vz}}(t)$ [L T^{-1}]: $\frac{dS_{\text{vz}}}{dt} = J_{\text{vz}}(t) - Q_{\text{vz}}(t) - ET(t)$. Reflecting the influence of capillary forces

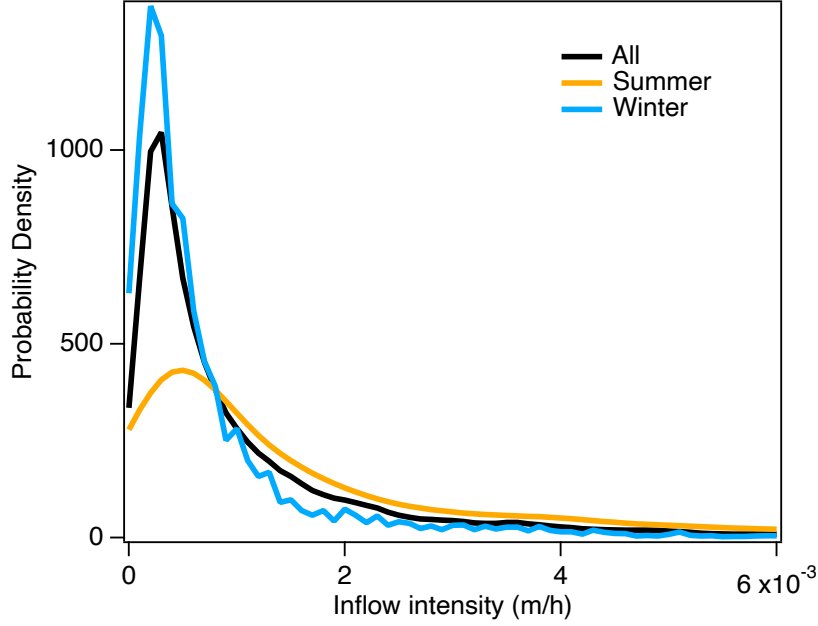


Fig. B3 Probability distributions for inflow intensities for summer (June, July, August), winter (December, January, February) and all days of the study period. During summer, probability for high intensity ($> 2 \times 10^{-3}$ m/h) inflows is greater (yellow curve) than during winter (blue curve), suggesting higher runoff generation (and consequently lower infiltration) during summer months.

on transiently saturated flow, the discharge-storage relationship for the vadose zone is presumed to be non-linear [66, 78]: $Q_{vz}(t) = K_{sat} \left(\frac{S_{vz}(t)}{S_{vz,max}} \right)^g$, where K_{sat} [$L T^{-1}$], $S_{vz,max}$ [L] and g [-] represent, respectively, the saturated hydraulic conductivity of the vadose zone, maximum depth of the vadose zone, and a power-law exponent. Evapotranspiration was calculated from the product of vadose zone saturation and hourly estimates of potential evapotranspiration, $PET(t)$ [$L T^{-1}$], obtained from the NASA Land Surface Model [78]: $ET(t) = \frac{S_{vz}(t)}{S_{vz,max}} PET(t)$. A fixed fraction γ_2 of outflow from the vadose zone infiltrates to groundwater, $J_{gw}(t) = \gamma_2 Q_{vz}(t)$ [$L T^{-1}$], while the rest flows to the stream as interflow, $Q_{int}(t) = (1 - \gamma_2) Q_{vz}(t)$ [$L T^{-1}$].

Water entering groundwater, $J_{gw}(t)$ [$L T^{-1}$] can accumulate with time, $\frac{dS_{gw}}{dt}$ [$L T^{-1}$], or flow to the stream, $Q_{gw}(t)$ [$L T^{-1}$]: $\frac{dS_{gw}}{dt} = J_{gw}(t) - Q_{gw}(t)$. Groundwater storage is divided into an active portion that influences stream flow, $S_{gw,a}(t)$ [L], and a fixed passive portion, $S_{gw,p}$ [L], that does not influence stream flow but can store and release solutes [18]: $S_{gw}(t) = S_{gw,a}(t) + S_{gw,p}$. The groundwater discharge-storage relationship depends linearly on active storage [79]: $Q_{gw}(t) = k_{gw} S_{gw,a}(t)$ where k_{gw} [T^{-1}] is a response rate constant for the active portion of groundwater storage. The coupled set of differential equations for storage in the vadose zone and groundwater were numerically integrated within Wolfram Mathematica (v. 12.3).

837 Appendix C Transient-Transit Time Distribution 838 (T-TTD) Model

839 T-TTD theory was used to estimate, on an hourly basis, the mean age of water in the
840 stream and the fraction of stream flow derived originally from snow melt, assuming
841 that outflow from the vadose zone and groundwater is selected uniformly by age;
842 i.e., a uniform storage selection function was adopted for both the vadose zone and
843 groundwater control volumes (Figure B2) [35, 80]. These calculations are described in
844 turn.

845 C.1 Deriving Solutions for the Mean Age of Water

846 The mean age of water in the stream, $\mu_{\text{stream}}(t)$ [T], was calculated from the expected
847 value of the stream's age distribution, $\mu_{\text{stream}}(t) = \int_0^\infty T p_{\text{stream}}(T, t) dT$, where the
848 variable T represents water age and $p_{\text{stream}}(T, t)$ [T⁻¹] is the probability density func-
849 tion (PDF) form of the stream's age distribution at any time t . The latter can be
850 expressed as the flow-weighted sum of age distributions in runoff, interflow and ground-
851 water discharged to the stream: $p_{\text{stream}}(T, t) = \omega_{\text{run}}(t) p_{\text{run}}(T, t) + \omega_{\text{int}}(t) p_{\text{int}}(T, t) +$
852 $\omega_{\text{gw}}(t) p_{\text{gw}}(T, t)$. The weighting functions are defined as follows:

$$\omega_{\text{run}}(t) = \frac{Q_{\text{run}}(t)}{Q_{\text{stream}}(t)}, \omega_{\text{int}}(t) = \frac{Q_{\text{int}}(t)}{Q_{\text{stream}}(t)}, \omega_{\text{gw}}(t) = \frac{Q_{\text{gw}}(t)}{Q_{\text{stream}}(t)}, \quad (\text{C1})$$

853 Combining these equations we arrive at the following expression for the mean age
854 of water in the stream at any time t , where $\mu_{\text{run}}(t)$, $\mu_{\text{int}}(t)$, and $\mu_{\text{gw}}(t)$ are the expected
855 values of water age in direct runoff, interflow and groundwater, respectively:

$$\mu_{\text{stream}}(t) = \omega_{\text{run}}(t) \mu_{\text{run}}(t) + \omega_{\text{int}}(t) \mu_{\text{int}}(t) + \omega_{\text{gw}}(t) \mu_{\text{gw}}(t) \quad (\text{C2})$$

856 For this analysis we assumed that runoff has a single age, T_{run} [T], when it is
857 reaches the stream: $p_{\text{run}}(T, t) = \delta(T - T_{\text{run}})$, where δ is the Dirac Delta function.
858 Therefore, the mean age of runoff entering the stream is $\mu_{\text{run}}(t) = T_{\text{run}}$.

859 The formulae used to estimate the mean age of water in interflow and groundwater
860 were derived by first constructing a generic solution for the mean age of water leaving
861 a control volume assuming that: (1) inflow has some prescribed age distribution which
862 can vary with time; (2) outflow from the control volume occurs by both discharge
863 and evapotranspiration; and (3) water leaving the control volume by discharge and
864 evapotranspiration is sampled from storage uniformly by age. This generic solution
865 for the mean age of water in, and leaving, a control volume was then tailored to yield
866 separate formulae for the mean age of water entering the stream from the vadose zone
867 and groundwater.

868 C.2 Mean Age of Water Leaving a Generic Control Volume

869 Consider a control volume drawn around either the vadose zone or groundwater (see
870 below) that receives inflow at rate, $J(t)$ [L T⁻¹], and loses water through discharge
871 at rate, $Q(t)$ [L T⁻¹], and evapotranspiration at rate, $ET(t)$ [L T⁻¹] (all volumes and
872 fluxes are normalized by the catchment area). From volume balance, the change in

873 water stored in the control volume, $S(t)$ [L], is therefore equal to the instantaneous
874 difference between inflows and outflows: $\frac{dS}{dt} = J(t) - Q(t) - ET(t)$. We make the fol-
875 lowing additional assumptions: (1) water entering the control volume with inflow $J(t)$
876 has an age distribution that varies with time, $P_{in}(T, t)$ [-], where the variable T rep-
877 represents the random variable for water age, and the age distribution is expressed as a
878 cumulative distribution function (CDF); and (2) water leaving the control volume by
879 either discharge $Q(t)$ or evapotranspiration $ET(t)$ is uniformly sampled from water in
880 storage (i.e., a uniform StorAge Selection (SAS) function is adopted for both discharge
881 and evapotranspiration). Under such conditions, the time evolution of the age distri-
882 bution of water stored in the control volume, $P(T, t)$ [-], expressed here as a CDF, is
883 governed by the following Age Conservation Equation (ACE) [37]:

$$\frac{\partial S(t)P(T, t)}{\partial t} = J(t)P_{in}(T, t) - ET(t)P(T, t) - Q(t)P(T, t) - S(t)\frac{\partial P(T, t)}{\partial T} \quad (C3a)$$

$$P(T = 0, t) = 0 \quad (C3b)$$

$$P(T, t = 0) = H(T - T_0) \quad (C3c)$$

$$H(x) = \begin{cases} 0, & x < 0 \\ 1, & x \geq 0 \end{cases} \quad (C3d)$$

884 In this equation, the time rate of change in the fraction of water in storage with
885 age less than or equal to T (left hand side) depends on the inflow of water of age
886 $P_{in}(T, t)$ (first term on right hand side), outflow of water by evapotranspiration and
887 discharge under uniform selection (second and third terms), and the aging of water in
888 storage with time (fourth term). The boundary condition (Equation C3b) stipulates
889 that there is no water in storage with age less than or equal to $T = 0$, while the initial
890 condition (Equation C3c) stipulates that water in storage at time $t = 0$ has a single
891 age equal to $T = T_0$, where $H(\cdot)$ is the unit step or Heaviside function (Equation
892 C3d). Expanding the left hand side of the equation and substituting the conservation
893 equation for water in storage we obtain the following simplified form of the ACE:

$$S(t)\frac{\partial P(T, t)}{\partial t} = J(t)(P_{in}(T, t) - P(T, t)) - S(t)\frac{\partial P(T, t)}{\partial T} \quad (C4)$$

894 The ACE can also be expressed in terms of probability density functions (PDFs)
895 for the age distributions of water leaving and entering the control volume ($p(T, t)$ and
896 $p_{in}(T, t)$, respectively), by taking the derivative of both sides of Equation C4 with
897 respect to the age variable T , where $p(T, t) = \frac{\partial P(T, t)}{\partial T}$ and $p_{in}(T, t) = \frac{\partial P_{in}(T, t)}{\partial T}$:

$$S(t)\frac{\partial p(T, t)}{\partial t} = J(t)(p_{in}(T, t) - p(T, t)) - S(t)\frac{\partial p(T, t)}{\partial T} \quad (C5a)$$

$$p(T = 0, t) = 0 \quad (C5b)$$

$$p(T, t = 0) = \delta(T - T_0) \quad (C5c)$$

898 The PDF form of the boundary and initial conditions (Equations C5b and C5c),
 899 follows by differentiating Equations C3b and C3c with respect to the age variable T .
 900 Taking the expected value, $\int_0^\infty T(\cdot) dT$, of each term in the PDF form of the ACE
 901 along with its boundary and initial conditions (Equations C5a-C5c) yields, after some
 902 manipulation, the following differential equation for the mean age of water in storage:

$$\frac{d\mu(t)}{dt} = \frac{J(t)}{S(t)}(\mu_{in}(t) - \mu(t)) + 1, \quad S(t) > 0 \quad (C6a)$$

$$\mu(t=0) = T_0 \quad (C6b)$$

903 Here, $\mu(t)$ represents the mean age (or expected value) of the water in and
 904 leaving the control volume at any time t , and $\mu_{in}(t)$ represents the mean age (or
 905 expected value) of water entering the control volume with inflow at any time t :
 906 $\mu(t) = \int_0^\infty T p(T, t) dT$ and $\mu_{in}(t) = \int_0^\infty T p_{in}(T, t) dT$. The above differential equation
 907 can be solved to yield the following exact solution for the mean age of water discharged
 908 from the control volume under uniform sampling:

$$\mu(t) = T_0 e^{-\tau_J(t)} + \int_0^t e^{-(\tau_J(t)-\tau_J(x))} \left(1 + \frac{J(x)}{S(x)} \mu_{in}(x) \right) dx, \quad S(t) > 0 \quad (C7a)$$

$$\tau_J(t) = \int_0^t \frac{J(\nu)}{S(\nu)} d\nu \quad (C7b)$$

909 In the event that the water coming into the tank has a mean age of $\mu_{in} = 0$, the
 910 above solution simplifies as follows:

$$\mu(t) = T_0 e^{-\tau_J(t)} + \int_0^t e^{-(\tau_J(t)-\tau_J(x))} dx, \quad \mu_{in}(t) = 0, \quad S(t) > 0 \quad (C8a)$$

911 C.3 Mean Age of Water in Interflow

912 From the generic solution presented above, the mean age of water leaving the vadose
 913 zone and entering the stream as interflow can be written as follows, where we have
 914 assumed the initial age of water in the vadoze zone at time $t = 0$ is, $T_{0,vz} = 0$:

$$\mu_{int}(t) = \int_0^t e^{-(\tau_{J,vz}(t)-\tau_{J,vz}(x))} dx, \quad S_{vz}(t) > 0 \quad (C9a)$$

$$\tau_{J,vz}(t) = \int_0^t \frac{J_{vz}(\nu)}{S_{vz}(\nu)} d\nu \quad (C9b)$$

915 C.4 Mean Age of Water in Groundwater

916 Accounting for the fact that water entering the groundwater (from the vadose zone)
 917 has some time varying mean age, $\mu_{int}(t)$, and building on the results presented above,
 918 we can write down the following solution for the mean age of water leaving the
 919 groundwater and entering the stream:

$$\mu_{gw}(t) = T_{0,gw} e^{-\tau_{J,gw}(t)} + \int_0^t e^{-(\tau_{J,gw}(t) - \tau_{J,gw}(x))} \left(1 + \frac{J_{gw}(x)}{S_{gw,a}(x) + S_{gw,p}} \mu_{int}(x) \right) dx, \quad S_{gw}(t) > 0 \quad (C10a)$$

$$\tau_{J,gw}(t) = \int_0^t \frac{J_{gw}(\nu)}{S_{gw,a}(\nu) + S_{gw,p}} d\nu \quad (C10b)$$

920 Implementation of this last solution requires that we specify a value for the pas-
 921 sive groundwater depth, $S_{gw,p}$. All else being equal, the model-predicted mean age of
 922 groundwater will increase as the depth of the passive groundwater is increased. Sub-
 923 surface flow discharged to streams in the Mesozoic Lowland hydrogeomorphic region
 924 (HGMR, where Flatlick Branch is located, see Figure A1b) is much younger than that
 925 in other Chesapeake Bay HGMRs (Figure C4). For the Flatlick Branch catchment,
 926 the median age of subsurface flow discharged to stream is 5.2 years [38] (right panel
 927 in Figure C4).

928 To identify the passive groundwater depth that would yield a similar model-
 929 predicted median subsurface flow age, we conducted the following numerical exper-
 930 iment. First, to provide sufficient time for the mean age of groundwater to reach a
 931 dynamic equilibrium, we created a 30-year synthetic time series of hourly precipi-
 932 tation, temperature and PET (the inputs to our HBV and Hydrologic models), by
 933 repeating the measured ten-year hourly time series for these three climate variables at
 934 our site a total of three times. Second, using the parameter values for the top-ranked
 935 model (see Table 1) we solved the HBV and Hydrologic models (see Sections A and
 936 B) for the entire 30-year period, yielding hourly estimates for all stocks and flows of
 937 water through the catchment. Third, from the 30-year time series of catchment stocks
 938 and flows we computed hourly estimates for the mean ages of water discharged to
 939 the stream from the vadose zone and groundwater (from equations (C9a) and (C10a),
 940 respectively). This allowed us to calculate the mean age of subsurface flow using the
 941 following equation:

$$\mu_{sub}(t) = \frac{Q_{int}(t)}{Q_{int}(t) + Q_{gw}(t)} \mu_{int}(t) + \frac{Q_{gw}(t)}{Q_{int}(t) + Q_{gw}(t)} \mu_{gw}(t) \quad (C11)$$

942 Finally, we computed the median subsurface flow age from the last six years of
 943 hourly mean subsurface flow age simulations. The last two steps were then repeated
 944 for 5 different choices of the passive groundwater depth, ranging from 0 to 1.5 m.
 945 The results of this numerical experiment, which are presented in Figure C5, indicate
 946 that the median groundwater age increases linearly with passive groundwater storage.

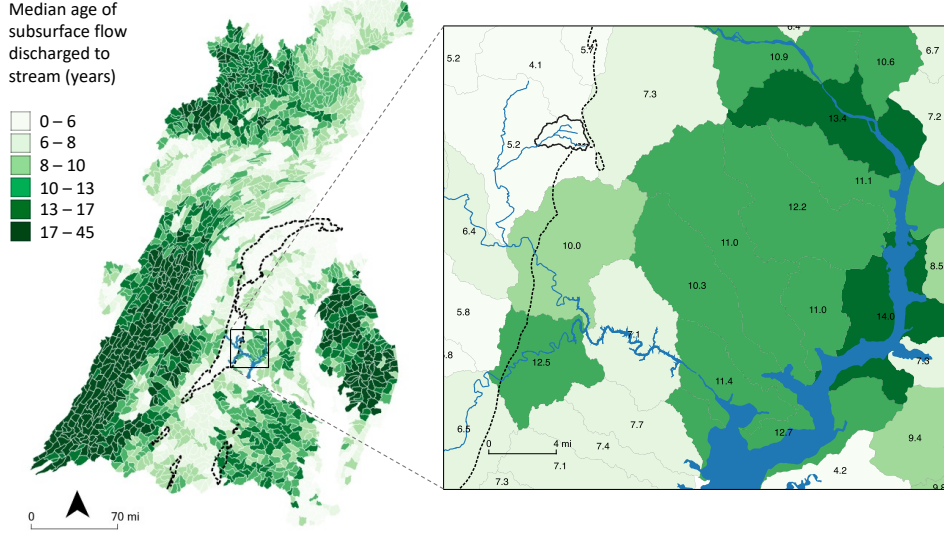


Fig. C4 Median age of subsurface flow discharged to streams in different sub-watersheds within the Chesapeake Bay watershed. Thick dashed curve denotes the Mesozoic Lowland HGMR. Solid black line (right panel) denotes the Flatlick Branch catchment.

From this plot we infer that, for the Flatlick Branch site, a median subsurface flow age of 5.2 years corresponds to a passive groundwater depth of $S_{gw,p} \approx 1.4\text{m}$. This choice of passive groundwater depth was therefore adopted for all age simulations described in the main text. We also adopted $T_{0,vz} = 0$ years and $T_{0,gw} = 15$ years for the initial ages of water in the vadose zone and groundwater, respectively.

C.5 Snow Melt Fraction

In Figure B2b, the variable, $C_{\text{stream}}(t) \in [0, 1]$ [-], represents the fraction of stream flow, at any time t , that originated as snow melt. The end members, $C_{\text{stream}}(t) = 1$ or $C_{\text{stream}}(t) = 0$, therefore represent conditions where the stream flow is 100% derived from snow melt or rainfall, respectively. As noted in the main text, we assumed that direct runoff has an effective age of $T_{\text{run}} = 0$. Given this assumption and for the choice of a uniform StorAge Selection (SAS) function for discharge from the vadose zone and groundwater (45), the fraction $C_{\text{stream}}(t)$ can be estimated directly from solute mass balance and the dynamic water balance calculations presented in Section B, as the flow weighted sum of the snow melt fraction of runoff, interflow, and groundwater ($C_{\text{run}}(t)$, $C_{\text{int}}(t)$, $C_{\text{gw}}(t)$, respectively):

$$C_{\text{stream}}(t) = \omega_{\text{run}}(t)C_{\text{run}}(t) + \omega_{\text{int}}(t)C_{\text{int}}(t) + \omega_{\text{gw}}(t)C_{\text{gw}}(t) \quad (\text{C12a})$$

$$C_{\text{run}}(t) = C_{J,vz}(t) = \frac{s(t)}{s(t) + r(t)} \quad (\text{C12b})$$

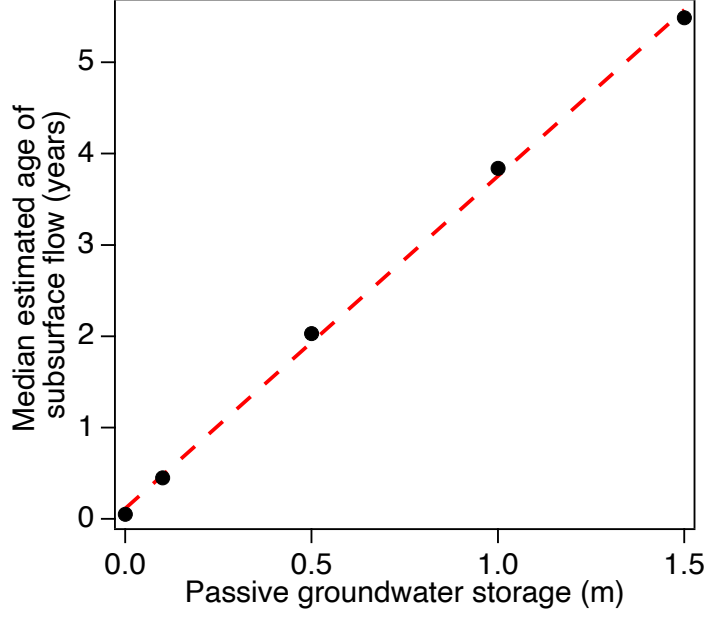


Fig. C5 Model-predicted median subsurface flow age discharged to Flatlick Branch for different choices of passive groundwater storage depth.

$$C_{\text{int}}(t) = C_{J, gw}(t) = \frac{1}{S_{vz}(t)} \int_0^t C_{J, vz}(t_i) J_{vz}(t_i) e^{-(\tau_{Q, vz}(t) - \tau_{Q, vz}(t_i))} dt_i \quad (\text{C12c})$$

$$C_{gw}(t) = \frac{1}{S_{gw, a}(t) + S_{gw, p}} \int_0^t C_{J, gw}(t_i) J_{gw}(t_i) e^{-(\tau_{Q, gw}(t) - \tau_{Q, gw}(t_i))} dt_i \quad (\text{C12d})$$

$$\tau_{Q, vz}(t) = \int_0^t \frac{Q_{vz}(r) + ET(r)}{S_{vz}(r)} dr \quad (\text{C12e})$$

$$\tau_{Q, gw}(t) = \int_0^t \frac{Q_{gw}(r)}{S_{gw, a}(r) + S_{gw, p}} dr \quad (\text{C12f})$$

963 The expression for $C_{\text{run}}(t)$ follows from a mass balance over rainfall and snow melt
 964 entering the catchment with snow melt fractions of $C_{\text{rain}} = 0$ and $C_{\text{snow}} = 1$, respec-
 965 tively (Equation C12b and top of Figure B2b). The integrals appearing in Equations
 966 C12c and C12d represent the sum over all snow melt entering the vadose zone and
 967 groundwater control volumes, respectively, from time $t_i \in [0, t]$. The exponential terms
 968 appearing in these integrals account for the outflow of snow melt under uniform sam-
 969 pling, the rate of which depends on outflow-weighted time $\tau_Q(t)$ [-]. The latter takes on
 970 different functional forms for the vadose zone (where snow melt can leave the control
 971 volume by both discharge and evapotranspiration, Equation C12e) and groundwater
 972 (where snow melt leaves the control volume by discharge alone, Equation C12f). As
 973 described above, for these calculations we set the volume of passive groundwater to
 974 $S_{gw, p} = 1.4\text{m}$.

975 Appendix D Model Calibration

976 D.1 Data Sources

977 HBV and hydrologic model predictions of catchment hydrology are driven by hourly
978 measurements of precipitation, air temperature, and potential evapotranspiration
979 (PET) (Figure B2a and Figure 1a). Ten years (water years 2008-2017) of measured
980 hourly precipitation (with a heated gage) and air temperature at Dulles Interna-
981 tional Airport (station ID WBAN:93738, located approximately 6 km north of the
982 Flatlick monitoring station, see Figure A1a) were downloaded from the National
983 Oceanic and Atmospheric Administration website (www.ncei.noaa.gov). Hourly esti-
984 mates of local PET (for a pixel $(0.125^\circ \times 0.125^\circ)$ centered on the nearby Occoquan
985 Forest) were obtained from NASA’s land surface model (NLDAS-Noah) using the
986 Data Rods Explorer (apps.hydroshare.org/apps/data-rods-explorer). Daily ground-
987 water elevation (in feet to groundwater from land surface, for comparison with
988 model-predictions of active groundwater depth, see Figure J10) were obtained for
989 USGS well (#385638077220101), located approximately 10 km northeast of the USGS
990 stream monitoring station on Flatlick Branch (Figure A1). The hydrologic model was
991 calibrated and validated with ten-years (water years 2008-2017) of hourly stream flow
992 measurements at a USGS monitoring station on Flatlick Branch (USGS Station ID
993 01656903 Figure A1). T-TDD predictions of mean stream water age and snow melt
994 fraction (Section C) were compared with hourly in situ measurements of specific con-
995 ductance, which was collected at the same location on Flatlick Branch (USGS Station
996 ID 01656903), and over the same ten-year time period (water years 2008-2017) as the
997 stream flow measurements described above.

998 D.2 Parameter Inference

999 The hydrological component of our modeling framework includes seven parameters,
1000 which were inferred from stream flow measurements at Flatlick Branch as described
1001 in Section 13. Preliminary evaluation indicated that the posterior distributions of
1002 all parameters were well constrained (see Figures 1 c,d), with the exception of the
1003 active groundwater response rate for which the posterior values were more uniformly
1004 distributed, a problem also noted by Bertuzzo et al. [79]. As noted in the main text,
1005 we fixed the active groundwater response rate to $k_{\text{gw}} = 0.11 \text{ day}^{-1}$, based on the slope
1006 of log-transformed measurements of stream flow at Flatlick Branch against time for
1007 the master recession curve (Figure D6). Posterior distributions were derived for the
1008 remaining six parameters conditioned on this single value for the active groundwater
1009 response rate.

1010 Appendix E Estimating Deicer Application Rates

1011 In Northern Virginia, winter maintenance responsibilities fall under the jurisdiction
1012 of both public sector agencies (e.g., on roads, parking lots, etc.) and private sector
1013 companies (e.g., on private properties) [64]. Virginia Department of Transportation
1014 (VDOT) guidelines for deicer application rates in this area (which are based on Salt
1015 Institute’s standard road salt (as NaCl) application rates) are keyed to winter-weather

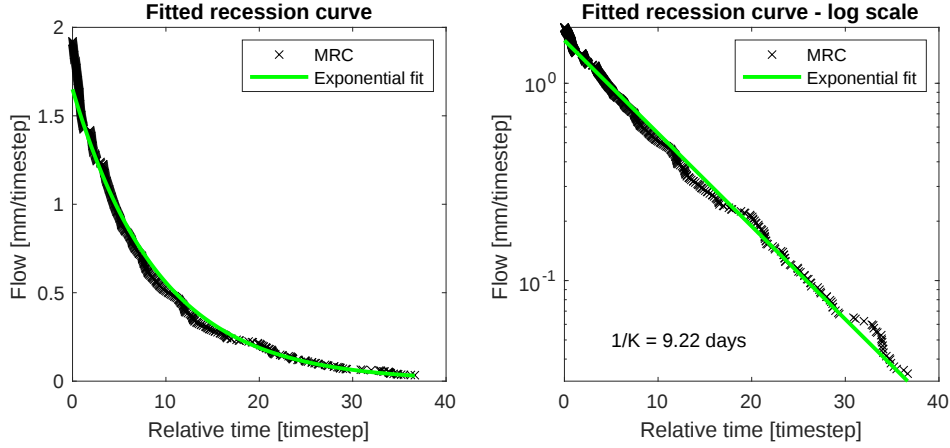


Fig. D6 Master recession curve (MRC) for the Flatlick Branch catchment.

mobilization levels, which range from a Mobilization Level 1 (corresponding to a recommended daily road salt application to impervious surfaces of 0.025 kg m^{-2}) to Mobilization Level 5 (corresponding to a recommended road salt application of 0.048 kg m^{-2}) (Table E2). The mobilization levels, in turn, depend on forecasts of precipitation probability, snow accumulation depth and ambient temperature (Table E2, see also Appendix D.a in Moltz et al. [64]). We re-constructed a ten-year timeseries of daily mobilization levels at Flatlick Branch based on historical weather data recorded at Dulles Airport (i.e., we used actual historical weather data as a proxy for historical weather forecasts, which were not available for the ten year period of interest here). To evaluate how well these simulated mobilization levels comport with reality, we obtained from VDOT the actual daily mobilization levels recorded for the Northern Virginia region during the winter seasons of 2019 and 2020. Mobilization levels predicted with measured climate data at Dulles (black bars Figure E7) compare favorably with actual mobilization levels recorded by VDOT (red bars Figure E7), although there are many cases where the actual mobilization level was higher than the predicted mobilization level, and more infrequently cases where the actual mobilization level was lower than the predicted mobilization level. These differences are due to weather forecast inaccuracies (e.g., that might predict precipitation when no precipitation in fact occurred) and the fact that the conditions listed for each mobilization level in Table E2 are just recommended guidelines; other factors, such as professional judgement [65], often influence the mobilization levels adopted on any particular winter day (personal communication, VDOT staff).

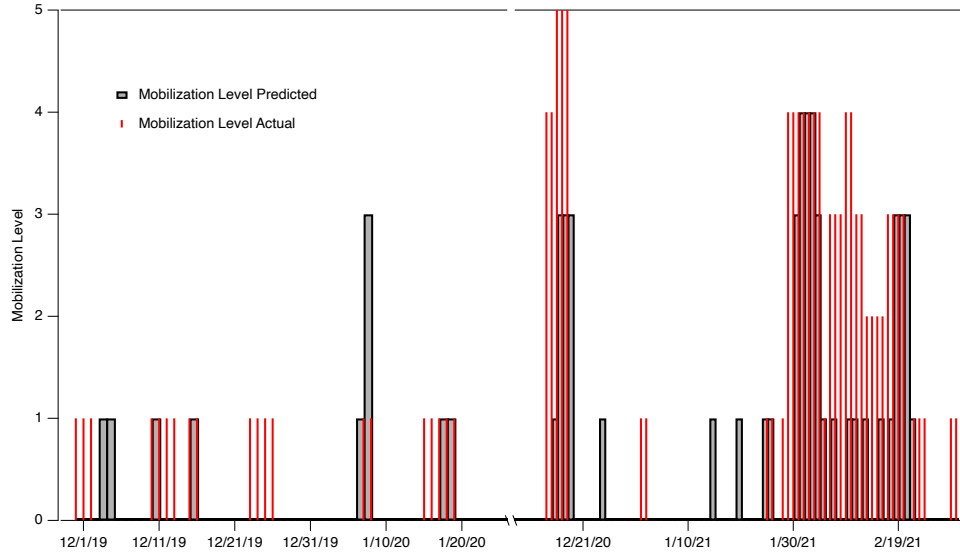


Fig. E7 Comparison of actual and model predicted winter weather mobilization levels in Northern Virginia for a two-year period.

Table E2 Salt Institute's standard road salt (as NaCl) application rates

Weather Forecast	Winter-Weather Mobilization Level	Road Salt (NaCl) Application Rate (kg/m ²)
P: >20%; S: possible; T: 30-36	Anti-ice	0.025
P: 20-49%; S: possible; T: 30-36	1	0.025
P: 50-100%; S: < 1 inch; T: 25-29	2	0.031
P: 50-100%; S: < 2 inch; T: 20-24	3	0.037
P: 50-100%; S: < 6 inch; T: 15-19	4	0.042
P: 50-100%; S: > 6 inch; T: 10-14	5	0.047

P: Chance of precipitation; S: Snowfall; T: Temperature (°F).

Appendix F Inferring Chloride Concentrations from Specific Conductance

To contextualize the specific conductance (SC) measurements on Flatlick Branch, we estimated from the SC measurements stream chloride concentrations which can, in turn, be directly compared to the U.S. Environmental Protection Agency's 1988 Ambient Water Quality Criteria for stream chloride concentration (chronic and acute thresholds of 230 and 860 mg L⁻¹, respectively) (EPA 1988). To convert SC measurements into chloride concentrations we utilized USGS measurements of SC and chloride on 27 grab samples collected at our field site on a monthly basis for the past three calendar years (2020-2022). Regressing log-transformed chloride against log-transformed

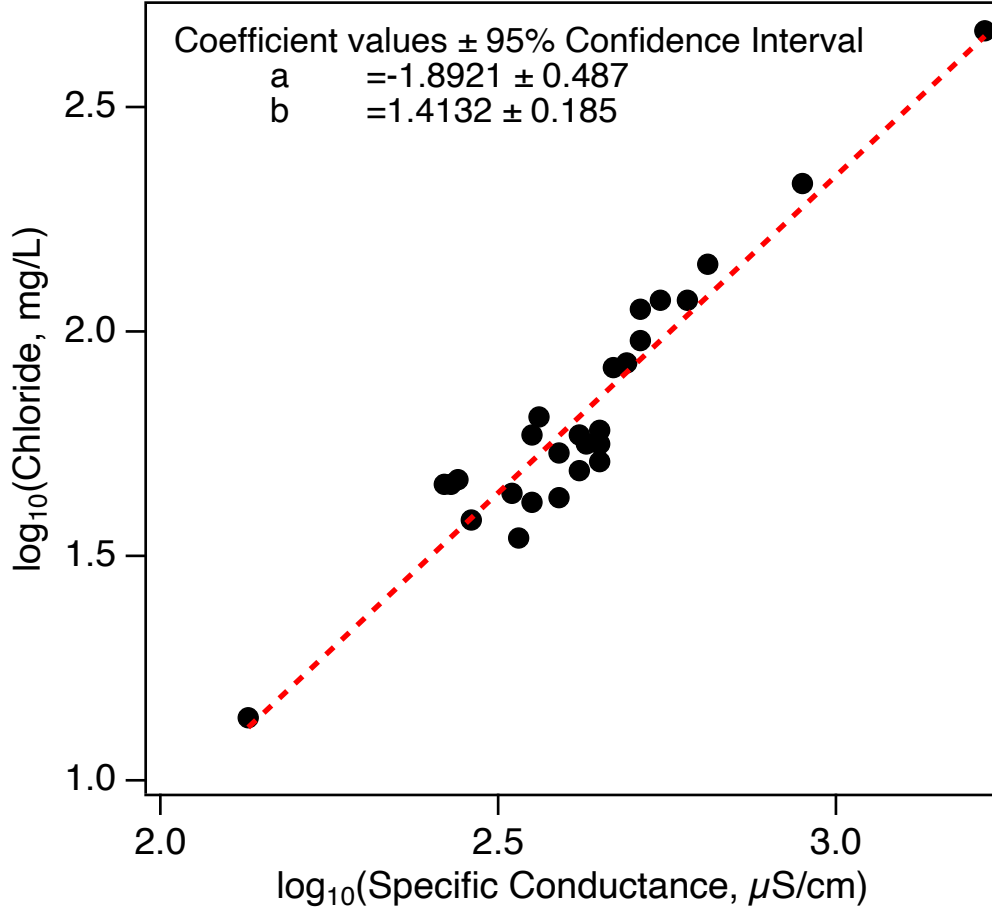


Fig. F8 Measured chloride vs in-stream specific conductance at Flatlick Branch.

SC we obtain the following power-law relationship between the two: $[\text{Cl}] = 10^a \times \text{SC}^b$, where $a = -1.89 \pm 0.48$ and $b = 1.41 \pm 0.18$ ($R^2=0.91$) (Figure F8).

Appendix G Deicer Application Rates and Stream Salinity

Analysis of salinity measurements on Flatlick Branch reveals a cause-and-effect relationship between peak stream salinity and rough estimates for the cumulative area-normalized mass of road salt deposited in the lead up to a snow melt event (Figure G9). The latter was estimated by first generating a daily time series of recommended salt application (in units of kg m^{-2}) based on local deicer application guidelines [64] and historical weather data (Section E). Over the eight year calibration and validation periods, we then summed the predicted daily mass of road salt (as sodium chloride)

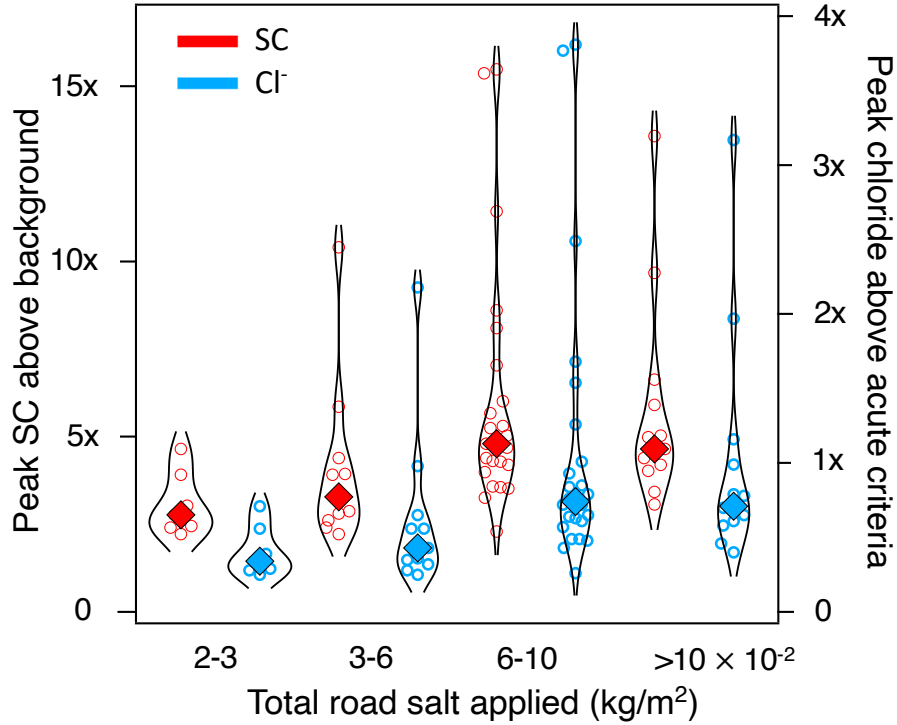


Fig. G9 Cause-and-effect relationship at Flatlick Branch between the cumulative road salt applied in the lead-up to a storm, and the peak stream SC and inferred chloride concentrations during snow melt events.

1059 applied from the end of the penultimate storm to the snow melt event in question.
 1060 Measurements of peak stream salinity at Flatlick Branch increase non-linearly with
 1061 these predictions of cumulative deicer application, from at least two times the back-
 1062 ground level for 0.02 to 0.03 kg NaCl m⁻² to between three and fifteen times the
 1063 background level for > 0.1 kg NaCl m⁻² (red symbols in Figure G9). These specific
 1064 conductance values can be roughly translated into chloride concentrations using the
 1065 site-specific regression relationship described in Section F. At higher salt application
 1066 rates the inferred peak stream chloride concentrations are nearly four times the U.S.
 1067 EPA acute aquatic life threshold of 860 mg L⁻¹ (blue symbols in Figure G9).

1068 Appendix H Groundwater Specific Conductance in 1069 the Mesozoic Lowland HGMR

1070 We compiled historical data [42, 81] and data available thorough the National Ground-
 1071 water Monitoring Network (cida.usgs.gov/ngwmn), on specific conductance measured
 1072 in groundwater wells (depths ranging from 80-1000 feet) in the Mesozoic Lowland
 1073 hydrogeomorphic (HGMR) region of Virginia. Measured specific conductance in deep

groundwater ranged from 62-4500 uS/cm, with a median value of 542.5 uS/cm (Figure 3). These data are compiled in Tables H3-H5.

Table H3 Groundwater Specific Conductance for Wells in the Mesozoic Lowland HGMR Reported in Posner and Zenome, 1983

USGS Local Reference No.	Date	Depth of Well (feet)	Specific Conductance (uS/cm)
EC-10	11/27/79	858	970
51V-3E	7/30/75	100	780
51V-3E	7/30/75	520	750
51V-3E	7/30/75	900	4500
51V-24H	7/20/79	500	560
51V-24H	7/26/79	760	2000
50U-83C	6/24/80	-	2140
50U-84F	6/24/80	-	644
50T-16B	7/30/80	153	433
50S-2D	5/02/80	205	439
48R-3A	6/17/80	-	2710
DC-31	10/18/78	285	355
DB-47	10/18/78	600	275
51V-14F	10/24/78	423	295
51V-14F	10/25/88	880	290
51V-23H	7/09/79	1000	640
51V-5J	1/13/77	289	150
49U-11C	4/28/80	500	254
48R-4A	6/17/80	100	414
47R-11G	8/12/80	-	412

Appendix I Deconvolution Framework

As noted above, the T-TTD model produces time-varying estimates for the age-distribution of water in the stream, $p_{\text{stream}}(T, t)$ (units inverse age) where T is water age and t is time (units of age and time, respectively). The increment of salt mass discharged from the stream per unit time with age in the range T to $T + \Delta T$ at time t is therefore $\Delta \dot{M}_Q(T, t) = \dot{M}_Q(t) p_{\text{stream}}(T, t) \Delta T$. Here $\dot{M}_Q(t)$ (units of salt mass per time) is the measured salt mass discharged from the stream at time t (equal to measured salt concentration times measured stream discharge) and $\Delta \dot{M}_Q(T, t)$ is the portion of that discharged salt mass with age T at time t . The rate at which salt was added to the catchment (e.g., as deicers) at “injection” time $\tau = t - T$ is therefore the sum over all $\Delta \dot{M}_Q(T, t)$ for which the difference of t and T equals τ :

$$\dot{M}_J(\tau) = \sum_{\tau=t-T} \Delta \dot{M}_Q(T, t) \quad (\text{I13})$$

This can be visualized graphically by imagining a matrix in which the rows and columns represents time t and stream water age T , respectively, and the entries are our estimate of $\Delta \dot{M}_Q(T, t)$ obtained from the measured stream salt mass loading rate

Table H4 Groundwater Specific Conductance for Wells in the Mesozoic Lowland HGMR Reported in Froelich and Zenome, 1985

USGS Local Reference No.	Date	Depth of Well (feet)	Specific Conductance (uS/cm)
385320077251701	7/05/79	743	931
385320077251701	7/09/79	1000	538
384810077284501	4/18/77	320	600
384918077281201	4/17/77	290	437
385321077234001	12/21/76	309	160
385311077234501	1/13/77	289	150
385157077275301	4/20/77	400	398
385543077271901	8/18/75	860	1170
385617077271201	7/30/75	100	780
385617077271201	7/30/75	530	750
385617077271201	7/30/75	900	4500
385048077320001	4/20/77	-	585
385656077241201	10/04/74	-	215
390130077210801	10/04/74	-	320
390109077210701	10/04/74	-	188
385657077232701	11/12/74	-	200
385350077272501	7/20/79	500	547
385350077272501	7/26/79	500	2050
385205077275501	6/20/79	650	570
385205077275501	8/09/76	650	849

Table H5 Groundwater Specific Conductance for Wells in the Mesozoic Lowland HGMR in Virginia from the National Groundwater Monitoring Network (cida.usgs.gov/ngwmn)

USGS Local Reference No.	Date	Depth of Well (feet)	Specific Conductance (uS/cm)
385930000000000	7/22/03	120	551
385930000000000	7/22/03	120	507
385930000000000	2/03/05	120	330
385930000000000	2/03/05	120	432
385930000000000	5/05/05	120	482
385930000000000	5/05/05	120	445
385930000000000	8/01/05	120	520
385930000000000	8/01/05	120	523
385930000000000	8/07/07	120	599
385930000000000	8/07/07	120	606
385930000000000	7/28/09	120	698
385930000000000	7/28/09	120	712
385930000000000	8/17/11	120	795
385930000000000	8/17/11	120	784
385930000000000	8/12/15	120	977
385930000000000	8/12/15	120	946
384956000000000	7/23/03	80	62
384956000000000	7/23/03	80	66
384956000000000	7/28/15	80	72
384956000000000	7/28/15	80	74

1090 and our T-TTD theory estimates for the age distribution, $p_{\text{stream}}(T, t)$ (see above).
 1091 The salt mass loading rate to the catchment at time τ is simply the sum over all cells
 1092 in the matrix that fall along the line, $t = \tau + T$ [82, 83].

1093 Appendix J Additional Supplementary Figures

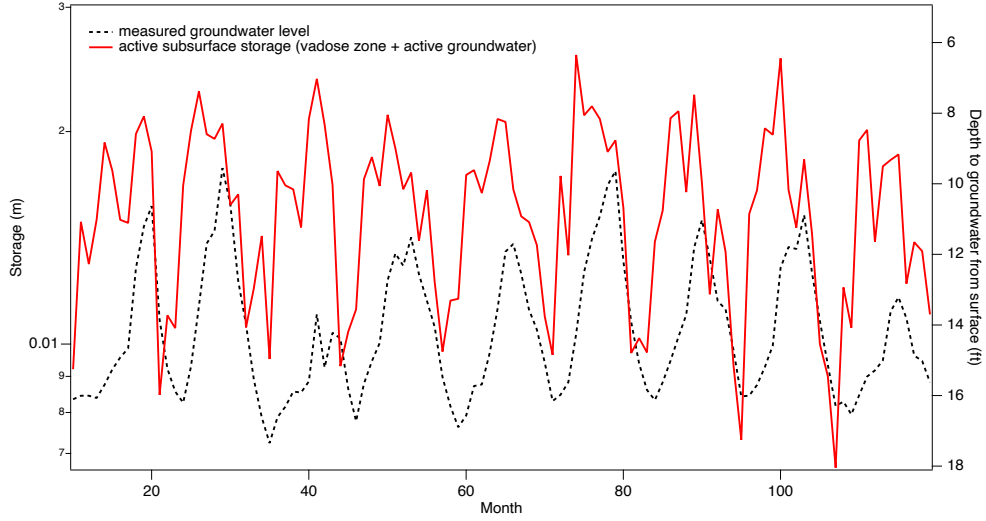


Fig. J10 Model-predicted storage (monthly averaged) in the vadose zone and groundwater (red curve) follows a seasonal pattern similar to that of the monthly groundwater level observed at a nearby USGS monitoring well (USGS well #385638077220101, see Figure A1)

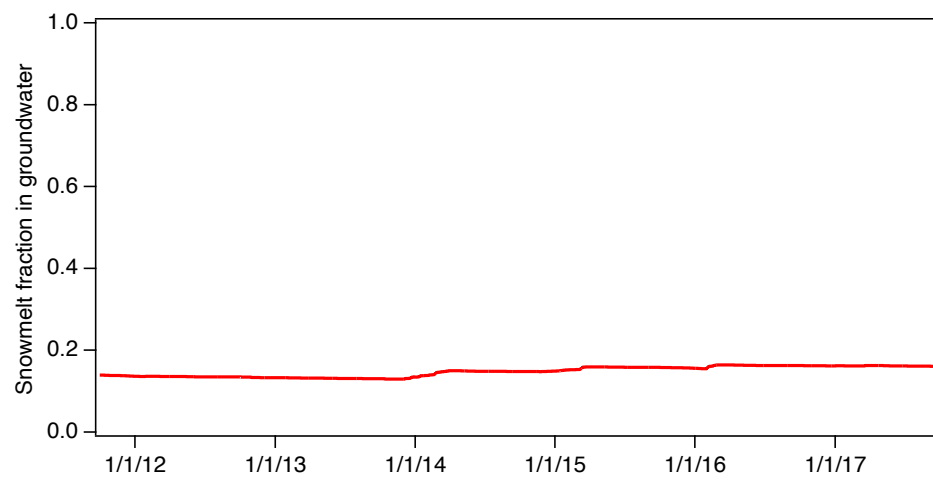


Fig. J11 Model-predicted fraction of snow melt in groundwater for the validation period.

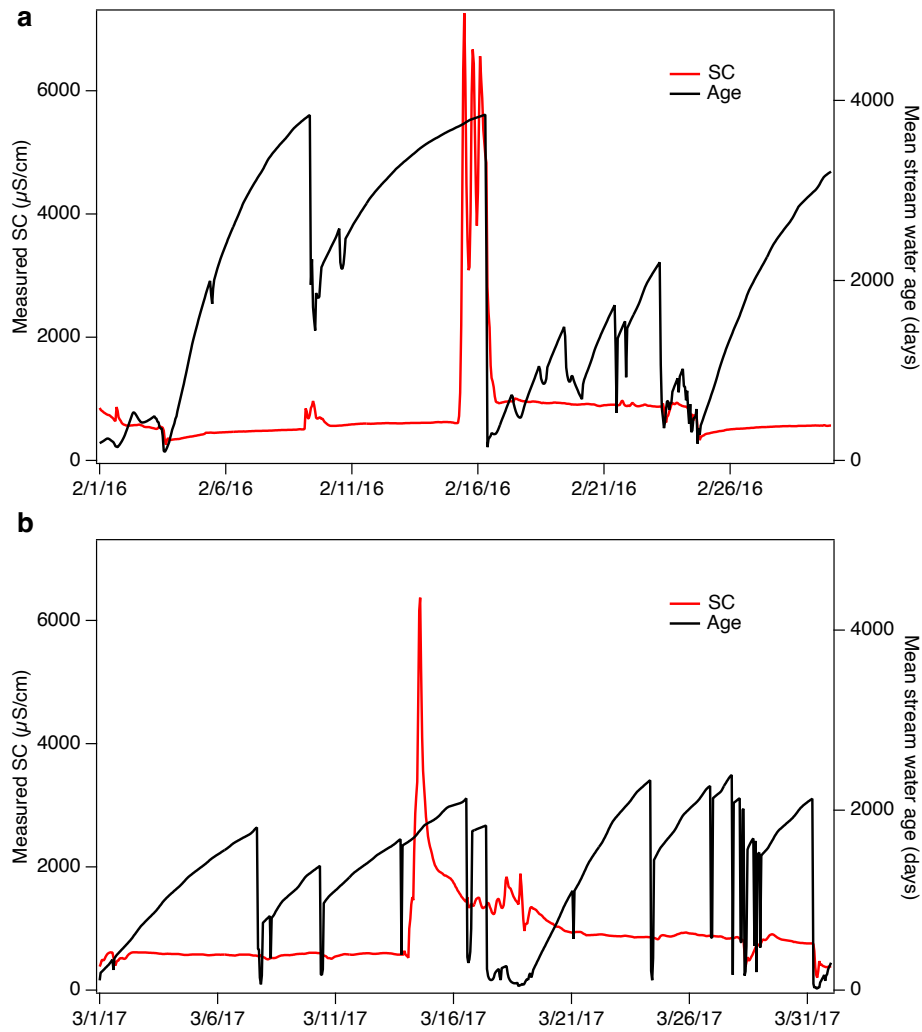


Fig. J12 Illustrative examples of periods when peaks in measured stream specific conductance (SC) do not perfectly align with reductions in the mean stream water age (reflecting contribution from younger direct runoff and/or interflow). SC pulses either arrive early (panel a) or are delayed (panel b) with respect to younger flows from direct runoff and/or interflow.