

**Supporting Information for**  
**“Surface flux and ocean heat transport convergence contributions to seasonal and interannual variations of ocean heat content”**

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**1 Heat conservation within an ocean layer of arbitrary depth**

To evaluate the drivers of ocean heat content we consider the temperature conservation equation integrated from the ocean free surface  $\eta$  to a depth  $D$ .

$$\int_{-D}^{\eta} \frac{\partial T}{\partial t} dz + \int_{-D}^{\eta} \nabla \cdot (\mathbf{u}T) dz = \int_{-D}^{\eta} \frac{1}{\rho_0 c_p} \frac{\partial Q}{\partial z} dz \quad (1)$$

where  $\mathbf{u}$  is the three-dimensional velocity field,  $T$  is ocean temperature,  $\rho_0$  is a reference density,  $c_p$  is the heat capacity of sea water, and  $Q$  represents the combined influence of radiation and turbulent fluxes of heat. Expanding the first term in (1) and accounting for time-variations in  $\eta$  and  $H$  (Leibniz integral rule) gives the following expression

$$\int_{-D}^{\eta} \frac{\partial T}{\partial t} dz = \frac{\partial}{\partial t} \int_{-D}^{\eta} T dz - T(\eta) \frac{\partial \eta}{\partial t} + T(-D) \frac{\partial(-D)}{\partial t} \quad (2)$$

Similarly, we can expand the second term in (1) into contributions from the horizontal and vertical circulation to give

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$$\int_{-D}^{\eta} \nabla \cdot (\mathbf{u}T) dz = \int_{-D}^{\eta} \nabla_h \cdot (\mathbf{u}_h T) dz + T(\eta) \frac{\partial \eta}{\partial t} - wT|^{-D} \quad (3)$$

where  $\mathbf{u}_h = (u, v)$  and we have noted that  $w(\eta) = \frac{\partial \eta}{\partial t}$  and  $wT|^{-D}$  indicates the product  $wT$  evaluated at  $-D$ . Finally, we can expand the right hand side of (1) to give

$$\int_{-D}^{\eta} \frac{1}{\rho_0 c_p} \frac{\partial Q}{\partial z} dz = \frac{1}{\rho_0 c_p} (Q_{net} - Q(-D)) \quad (4)$$

where  $Q_{net}$  is the net air-sea heat flux across the ocean-atmosphere interface. When  $D$  is fixed in time, equations (2)-(4) can be combined to give

$$\frac{\partial H}{\partial t} + \rho_0 c_p \left( \int_{-D}^{\eta} \nabla_h \cdot (\mathbf{u}_h T) dz - wT|^{-D} \right) = Q_{net} - Q(-D) \quad (5)$$

$$H = \rho_0 c_p \int_{-D}^{\eta} T dz \quad (6)$$

where  $H$  is ocean heat content integrated to a depth of  $D$ . In the case where  $D$  is the ocean bottom ( $D_b$ ), (5) can be simplified to

$$\frac{\partial H_{tot}}{\partial t} = C_{tot} + Q_{net} \quad (7)$$

$$C_{tot} = -\rho_0 c_p \int_{-D_b}^{\eta} \nabla_h \cdot (\mathbf{u}_h T) dz \quad (8)$$

where  $C_{tot}$  is the total heat transport convergence within an ocean column. When  $D$  is specified to be a time-invariant maximum climatological mixed layer depth ( $D_{mld}$ ; see main text for definition), turbulent mixing across the lower surface can be considered negligible such that  $Q(-D_{mld}) = 0$  resulting in

$$\frac{\partial H_{mld}}{\partial t} = C_{mld} + Q_{net} \quad (9)$$

$$C_{mld} = -\rho_0 c_p \left( \int_{-D_{mld}}^{\eta} \nabla_h \cdot (\mathbf{u}_h T) dz - wT|^{-D} \right) \quad (10)$$

where  $C_{mld}$  is the heat transport convergence within the layer bounded by the surface and  $-D_{mld}$ . Note that  $C_{mld}$  also includes the impact of vertical advection across  $-D_{mld}$ .

## 2 Application of a Kalman filter to quantify ocean heat transport divergences

Our approach is similar to that described by *Kelly et al.* [2014] but differs in a number of important ways. Firstly, we consider a near-global domain with a spatial resolution of  $1 \times 1$  degrees and focus on the drivers of local heat transport convergences, whereas *Kelly et al.* [2014] focus on meridional transports across zonal sections in the Atlantic. Secondly, we use multiple observation-based estimates of  $Q_{net}$  for  $H$  to estimate observational uncertainties. Lastly,  $\hat{C}$  is treated explicitly within the Kalman filter state transition matrix (along with  $\hat{Q}_{net}$ , and  $\hat{H}$ ) rather than inferred as an unknown control term. This last step is accomplished by including persistence forecasts of  $\hat{C}$  and  $\hat{Q}_{net}$  using values from the previous month. Uncertainties in the persistence approximation are estimated from an initial no-Kalman solution and act as a constraint on the variances of  $\hat{Q}_{net}$  and  $\hat{C}$ . By performing this step, we do not have to make any assumptions about the relative magnitude of variances in  $\hat{C}$  and  $\hat{Q}_{net}$ .

Kalman smoother estimates of  $\hat{H}_{mld}$  and  $\hat{Q}_{net}$  are shown along with smoothed observational estimates and their associated uncertainties in supplementary figure 1a-b for the same illustrative location as figure 2 in the main text. In addition, Kalman smoother estimates of  $\hat{C}_{mld}$  are compared with values calculated using a centered-difference approach without a Kalman filter (supplementary figure 1c). These comparisons demonstrate that, within our uncertainty estimates, predictions of  $\hat{H}_{mld}$  and  $\hat{Q}_{net}$  are consistent with the imposed observational constraints and that  $\hat{C}_{mld}$  is highly correlated with the equivalent value estimated as a residual without using a Kalman filter.

### 2.1 Forward models

In order to apply the Kalman filter, equations (7) and (9) are discretized using a centered difference approximation to give equations of the form

$$H_{t+1} = H_{t-1} + \Delta t \left[ \frac{1}{2} Q_{net,t+1} + Q_{net,t} + \frac{1}{2} Q_{net,t-1} + \frac{1}{2} C_{t+1} + C_t + \frac{1}{2} C_{t-1} \right] + u_H \quad (11)$$

where the subscript  $t$  is month and  $u_H$  is a normally distributed random variable with variance  $\sigma_H^2$  that represents the uncertainty in the forward model. In theory,  $u_H$  should be zero

as our forward model is an expression of heat conservation. In practice,  $u_H$  is specified to be a very small number for numerical stability. Forward predictions of  $C$  and  $Q_{net}$  are then estimated assuming persistence of values in the previous month.

$$Q_{net,t+1} = Q_{net,t} + u_Q \quad (12)$$

$$C_{t+1} = C_t + u_C \quad (13)$$

The uncertainties  $u_Q$  and  $u_C$  are specified using output from an initial no-Kalman solution and act as a prescribed constraint on the variance of our solutions for  $Q_{net}$  and  $C$ .

## 2.2 Kalman filter equations and state matrices

The equations and solutions to the Kalman RTS smoother are described in detail by *Wunsch* [1996]. In matrix form, our forward model is written

$$x_{t+1} = A_t x_t + \Gamma_t u_t \quad (14)$$

where  $x$  is a state vector,  $A$  the state transition matrix,  $\Gamma$  an identity matrix (as we assume that uncertainties in our forward model are uncorrelated), and  $u$  the uncertainties associated with our forward model. Observational constraints are introduced through the following relationship

$$E_t x_t + n_t = y_t \quad (15)$$

where  $E$  is the observation operator matrix,  $x$  the state vector,  $n$  the uncertainty in observations, and  $y$  an observation vector. For our model, these matrices take the following form

$$x_t = \begin{bmatrix} H_t & H_{t-1} & Q_{net,t+1} & Q_{net,t} & Q_{net,t-1} & C_{t+1} & C_t & C_{t-1} \end{bmatrix}^T \quad (16)$$

$$u_t = \begin{bmatrix} u_H & 0 & u_{Q_{net}} & 0 & 0 & u_C & 0 & 0 \end{bmatrix}^T \quad (17)$$

$$y_t = \begin{bmatrix} H_t^{obs} & Q_{net,t}^{obs} \end{bmatrix}^T \quad (18)$$

$$E_t = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \end{bmatrix} \quad (19)$$

$$A = \begin{bmatrix} 0 & 1 & \frac{\Delta t}{2} & \Delta t & \frac{\Delta t}{2} & \frac{\Delta t}{2} & \Delta t & \frac{\Delta t}{2} \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \quad (20)$$

### 3 Local Ekman forcing of heat content variability

Here, we derive an expression for the local impact of Ekman heat transport convergences ( $C_{ek}$ ) on ocean heat content variability. Following the same approach that *Gill and Niller* [1973] applied to Ekman density fluxes, we decompose Ekman heat transport convergences into two separate contributions: (i) an upper ocean contribution due to Ekman layer transports across horizontal temperature gradients ( $C_{ek}^{mld}$ ) and (ii) an ocean interior contribution due to Ekman pumping across vertical temperature gradients ( $C_{ek}^{int}$ ). Vertically integrating Ekman heat transports from the surface to the bottom of the Ekman layer ( $z_{ek}$ ) and noting that  $\nabla \cdot \mathbf{u}_{ek} = 0$  results in the following expression

$$C_{ek}^{mld} = -\rho_0 c_p \int_{-z_{ek}}^0 (\mathbf{u}_{ek} \cdot \nabla T) dz \quad (21)$$

If we also assume that vertical temperature gradients within the Ekman layer are negligible such that  $T(0) = T(-z_{ek})$ , this expression simplifies to

$$C_{ek}^{mld} = -\rho_0 c_p (\mathbf{U}_{ek} \cdot \nabla T_{ek}) \quad (22)$$

where  $\nabla T_{ek}$  is the horizontal gradient of temperature in the Ekman layer and

$$\mathbf{U}_{ek} = \left( \int_{-z_{ek}}^0 u_{ek} dz, \int_{-z_{ek}}^0 v_{ek} dz \right) = \frac{\boldsymbol{\tau} \times \hat{\mathbf{z}}}{\rho_o f} \quad (23)$$

where  $\boldsymbol{\tau} = (\tau_x, \tau_y)$  is wind stress at the ocean surface and  $f$  is the Coriolis parameter. In the ocean interior, Ekman pumping leads to the vertical displacement of isotherms that can be expressed as an equivalent heat flux,  $C_{ek}^{int}$ . We estimate this flux by integrating the heat transport convergence induced by vertical Ekman pumping velocities from the base of the Ekman layer,  $-z_{ek}$ , to the ocean floor,  $-D_b$ .

$$C_{ek}^{int} = -\rho_o c_p \int_{-D_b}^{-z_{ek}} w \frac{\partial T}{\partial z} dz \quad (24)$$

This expression can then be evaluated for the following boundary conditions

$$w(-z_{ek}) = \frac{1}{\rho_o} \text{curl} \left( \frac{\boldsymbol{\tau}}{f} \right) = w_{ek} \quad (25)$$

$$w(-D_b) = 0 \quad (26)$$

under the constraint that  $w$  changes linearly with depth such that

$$\frac{\partial w}{\partial z} = \frac{w_{ek}}{D_b - z_{ek}} \quad (27)$$

Expansion of equation (24) using an integration by parts and substitution of the boundary conditions then allows the following simplifications

$$C_{ek}^{int} = -\rho_o c_p \left( wT|_{-D_b}^{-z_{ek}} - \int_{-D_b}^{-z_{ek}} T \frac{\partial w}{\partial z} dz \right) \quad (28)$$

$$C_{ek}^{int} = -\rho_o c_p \left( w_{ek} \cdot T_{ek} - [D_b - z_{ek}] \cdot \bar{T} \cdot \frac{w_{ek}}{D_b - z_{ek}} \right) \quad (29)$$

$$\bar{T} = \frac{1}{D_b - z_{ek}} \int_{-D_b}^{-z_{ek}} T dz \quad (30)$$

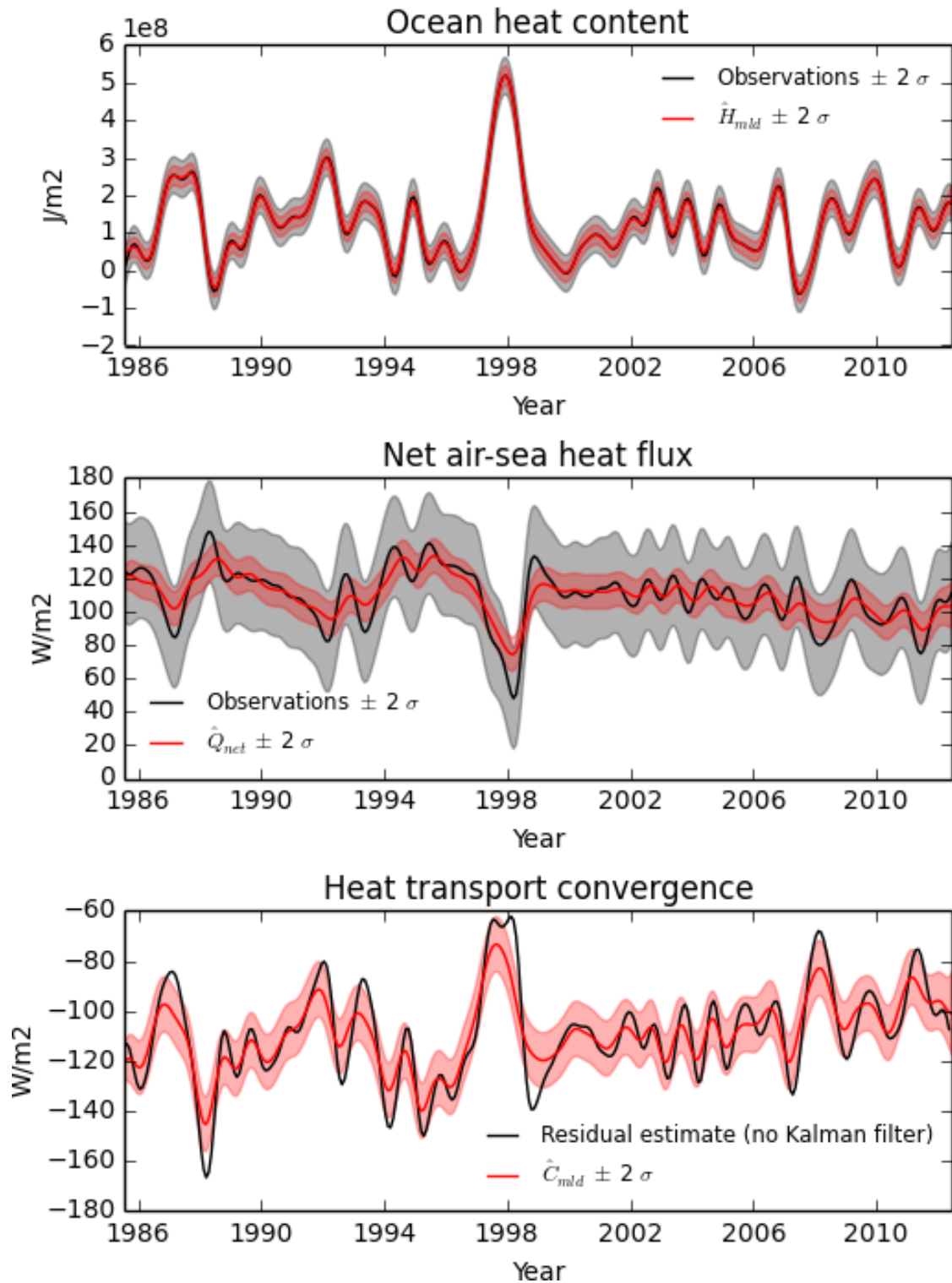
$$C_{ek}^{int} = -\rho_o c_p w_{ek} \cdot [T_{ek} - \bar{T}] \quad (31)$$

where  $T_{ek}$  is the temperature of the Ekman layer and  $\bar{T}$  is the depth-averaged temperature in the ocean interior. Combining equations (22) and (31) gives the following total expression for  $C_{ek}$

$$C_{ek} = C_{ek}^{mld} + C_{ek}^{int} = -\rho_0 c_p (\mathbf{U}_{ek} \cdot \nabla T_{ek} + w_{ek} \cdot [T_{ek} - \bar{T}]) \quad (32)$$

## References

- Gill, A., and P. Niller (1973), The theory of the seasonal variability in the ocean, *Deep Sea Research and Oceanographic Abstracts*, 20(2), 141–177.
- Kelly, K. A., L. Thompson, and J. Lyman (2014), The coherence and impact of meridional heat transport anomalies in the atlantic ocean inferred from observations\*, *Journal of Climate*, 27(4), 1469–1487.
- Wunsch, C. (1996), *The ocean circulation inverse problem*, Cambridge University Press.

Kalman smoother applied to  $0^{\circ}$  N,  $270^{\circ}$  E

**Figure 1.** (a-b) Kalman smoother estimates of  $\hat{H}_{mld}$  and  $\hat{Q}_{net}$  along with smoothed observations their associated uncertainties. (c) Kalman smoother estimate of  $\hat{C}_{mld}$  and estimated uncertainty compared with values calculated using a centred-difference approach without a Kalman filter.