We now focus on the following time-dependent PDE:

$$\mathcal{L}u(x,t) := \partial_t u(x,t) - \nabla \cdot (a(x)\nabla u(x,t)) = f(x,t), \quad x \in \Omega, \ t \in [0,T]$$
(1)

$$u(x,t) = 0,$$
 $x \in \partial\Omega, \ t \in [0,T]$ (2)

$$u(x,0) = u_0(x), \quad x \in \Omega$$
(3)

We will set up a prior on the solution u to the above problem. To do so we first let $v_h \in S_h$ be some approximation of the initial condition $u_0(x)$ in the FEM space S_h . To be more specific we will assume that $v_h(x) = \Phi(x)^* \gamma := \sum_{i=1}^J \phi_i(x) \gamma_i$. Note that $\Phi(x) := (\phi_1(x), \dots, \phi_J(x))^T$. We take the prior on u to be:

$$u \sim \mathcal{N}(m_0, V_0) \tag{4}$$

where $m_0(x,t) := v_h(x) = \Phi^*(x)\gamma$ (m_0 is constant in time). The prior covariance operator V_0 is defined as:

$$(V_0 g)(x,t) = \int_{\Omega} \int_0^T k_{x,t,y,s}^{(0)} g(y,s) ds dy$$
 (5)

where $k_{x,t,y,s}^{(0)}$ is defined as follows:

$$k_{x,t,y,s}^{(0)} := \sum_{i=1}^{J} \lambda_j \phi_j(x) \phi_j(y) k^{(0)}(t,s)$$
(6)

We now introduce a uniform time grid:

$$t_n = n\delta, \quad n = 0, 1, \dots, N$$

where δ is the spacing between consecutive times and $N = \frac{T}{\delta}$ (assume that N is an integer). The time kernel $k^{(0)}(t,s)$ will be taken to be:

$$k^{(0)}(t,s) := \sum_{i=0}^{N-1} (t - t_i)(s - t_i) \mathbb{1}_{(t_i, t_{i+1}]^2}(t,s)$$
(7)

We now introduce the following operators $\mathcal{I}_s := (I_1(s), \dots, I_J(s))^T$ where:

$$I_i(s)g := \int_{\Omega} \phi_i(x)g(x,s) dx \tag{8}$$

To move from $t = t_0 = 0$ to $t = t_1 = \delta$ we condition on observing $\mathcal{I}_{\delta}\mathcal{L}u = \mathcal{I}_{\delta}f =: F^{(1)}$. Let $\tilde{A}_{\delta} := \mathcal{I}_{\delta}\mathcal{L}$. For a fixed realisation of f (and so of $F^{(1)}$) we thus seek the following conditional distribution:

$$u|\{\tilde{A}_{\delta}u = F^{(1)}, f\} \sim \mathcal{N}(m_1, V_1)$$
 (9)

That this distribution is itself Gaussian follows from considering the following joint distribution:

$$\begin{pmatrix} u \\ \tilde{A}_{\delta}u \end{pmatrix} = \begin{pmatrix} I \\ \tilde{A}_{\delta} \end{pmatrix} u \sim \mathcal{N} \left(\begin{pmatrix} m_o \\ \tilde{A}_{\delta}m_0 \end{pmatrix}, \begin{pmatrix} V_0 & V_0\tilde{A}_{\delta}^* \\ \tilde{A}_{\delta}V_0 & \tilde{A}_{\delta}V_0\tilde{A}_{\delta}^* \end{pmatrix} \right)$$

It follows that the conditional distribution is Gaussian and the mean and covariance are given by:

$$m_1 = m_0 + V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} (F^{(1)} - \tilde{A}_{\delta} m_0)$$
(10)

$$V_1 = V_0 - V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} \tilde{A}_{\delta} V_0 \tag{11}$$

We now rewrite the mean update equation as follows:

$$m_1 = \left(1 - V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} \tilde{A}_{\delta}\right) m_0 + V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} F^{(1)}$$
(12)

To make progress we must now start computing various terms needed for our mean and covariance update rules. We start with $V_0 \tilde{A}_{\delta}^*$. We have:

$$V_0 \tilde{A}_{\delta}^* = V_0 \mathcal{L}^* (I_1(\delta)^*, \dots, I_J(\delta)^*)$$

We can thus see that we need to be able to compute terms of the form $V_0\mathcal{L}^*I_i(\delta)^* = V_0(I_i(\delta)\mathcal{L})^*$. Now since the operator $I_i(\delta)\mathcal{L}$ takes in a function on $\Omega \times [0,T]$ and outputs a real number its adjoint should take in a real number and output a function on $\Omega \times [0,T]$. This adjoint should satisfy the following relation:

$$\alpha(I_i(\delta)\mathcal{L}g) = \int_{\Omega} \int_0^T ((I_i(\delta)\mathcal{L})^*\alpha)(x,t)g(x,t)dtdx \ \forall g, \ \forall \alpha \in \mathbb{R}$$
 (13)

Using this we can now compute:

$$(V_0(I_i(\delta)\mathcal{L})^*\alpha)(x,s) = \int_{\Omega} \int_0^T k_{x,s,y,w}^{(0)}((I_i(\delta)\mathcal{L})^*\alpha)(y,w) dw dy$$
$$= \alpha(I_i(\delta)\mathcal{L}k_{x,s,\cdot,\cdot}^{(0)})$$
$$= \alpha \int_{\Omega} \phi_i(y)(\mathcal{L}k_{x,s,\cdot,\cdot}^{(0)})(y,\delta) dy$$

We now work out $(\mathcal{L}k_{x,s,...}^{(0)})(y,\delta)$ taking care to remember that x,s are fixed and so \mathcal{L} acts on the variables y,δ :

$$(\mathcal{L}k_{x,s,\cdot,\cdot}^{(0)})(y,\delta) = \partial_2 k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x) \phi_j(y) - k^{(0)}(s,\delta) \nabla_y \cdot \left(a(y) \nabla_y \sum_{j=1}^J (\lambda_j \phi_j(x) \phi_j(y)) \right)$$

$$= \partial_2 k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x) \phi_j(y) - k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x) \nabla_y \cdot (a(y) \nabla_y \phi_j(y))$$

So we can now compute:

$$(V_0(I_i(\delta)\mathcal{L})^*\alpha)(x,s) = \alpha \int_{\Omega} \phi_i(y)\partial_2 k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x)\phi_j(y) dy - \alpha \int_{\Omega} \phi_i(y)k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x)\nabla_y \cdot (a(y)\nabla_y \phi_j(y)) dy$$
$$= \alpha \partial_2 k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x)M_{ji} + \alpha k^{(0)}(s,\delta) \sum_{j=1}^J \lambda_j \phi_j(x)A_{ji}$$

where M is the Galerkin mass matrix and A is the Galerkin stiffness matrix, i.e. the matrices with entries given by:

$$M_{ij} := \int_{\Omega} \phi_i(x)\phi_j(x) dx \tag{14}$$

$$A_{ij} := \int_{\Omega} a(x) \nabla \phi_i(x) \nabla \phi_j(x) dx \tag{15}$$

Using this result we can deduce that:

$$(V_0 \tilde{A}_{\delta}^* \mathbf{v})(x, s) = \partial_2 k^{(0)}(s, \delta) \Phi(x)^* \Lambda M \mathbf{v} + k^{(0)}(s, \delta) \Phi(x)^* \Lambda A \mathbf{v}$$
(16)

for any $\mathbf{v} \in \mathbb{R}^J$, where $\Lambda = \operatorname{diag}\{\lambda_i\}_{i=1}^J$.

We now move onto computing:

$$\tilde{A}_{\delta}V_{0}\tilde{A}_{\delta}^{*} = \mathcal{I}_{\delta}\mathcal{L}V_{0}\mathcal{L}^{*}\mathcal{I}_{\delta}^{*}$$

$$= \begin{pmatrix} I_{1}(\delta) \\ \vdots \\ I_{J}(\delta) \end{pmatrix} \mathcal{L}V_{0}\mathcal{L}^{*} \begin{pmatrix} I_{1}(\delta)^{*} & \dots & I_{J}(\delta)^{*} \end{pmatrix}$$

This operator has ij-th entry which is given by:

$$\begin{split} &(\tilde{A}_{\delta}V_{0}\tilde{A}_{\delta}^{*})_{ij}\alpha = I_{i}(\delta)\mathcal{L}V_{0}\mathcal{L}^{*}I_{j}(\delta)^{*}\alpha \\ &= \int_{\Omega}\phi_{i}(x)\left[(\mathcal{L}V_{0}(I_{j}(\delta)\mathcal{L})^{*}\alpha)(x,\delta)\right]\mathrm{d}x \\ &= \int_{\Omega}\phi_{i}(x)\left[\alpha\partial_{1}\partial_{2}k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}\phi_{l}(x)M_{lj} + \alpha\partial_{1}k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}\phi_{l}(x)A_{lj} \\ &- \alpha\partial_{2}k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}M_{lj}\nabla\cdot(a(x)\nabla\phi_{l}(x)) - \alpha k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}A_{lj}\nabla\cdot(a(x)\nabla\phi_{l}(x))\right]\mathrm{d}x \\ &= \alpha\partial_{1}\partial_{2}k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}M_{il}M_{lj} + \alpha\partial_{1}k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}M_{il}A_{lj} \\ &+ \alpha\partial_{2}k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}M_{lj}A_{il} + \alpha k^{(0)}(\delta,\delta)\sum_{l=1}^{J}\lambda_{l}A_{il}A_{lj} \\ &= \alpha\partial_{1}\partial_{2}k^{(0)}(\delta,\delta)(M\Lambda M)_{ij} + \alpha\partial_{1}k^{(0)}(\delta,\delta)(M\Lambda A)_{ij} + \alpha\partial_{2}k^{(0)}(\delta,\delta)(A\Lambda M)_{ij} + \alpha k^{(0)}(\delta,\delta)(A\Lambda A)_{ij} \end{split}$$

We can thus conclude that $\tilde{A}_{\delta}V_0\tilde{A}_{\delta}^*$ is the $J\times J$ matrix given by:

$$\tilde{A}_{\delta}V_{0}\tilde{A}_{\delta}^{*} = \partial_{1}\partial_{2}k^{(0)}(\delta,\delta)M\Lambda M + \partial_{1}k^{(0)}(\delta,\delta)M\Lambda A + \partial_{2}k^{(0)}(\delta,\delta)A\Lambda M + k^{(0)}(\delta,\delta)A\Lambda A \tag{17}$$

We now give the various derivatives for our time kernel which are required (!!these should be checked rigorously!!):

$$\partial_1 k^{(0)}(t,s) = \sum_{i=0}^{N-1} (s-t_i) \mathbb{1}_{(t_i,t_{i+1}]^2}(t,s)$$

$$\partial_2 k^{(0)}(t,s) = \sum_{i=0}^{N-1} (t-t_i) \mathbb{1}_{(t_i,t_{i+1}]^2}(t,s)$$

$$\partial_1 \partial_2 k^{(0)}(t,s) = \sum_{i=0}^{N-1} \mathbb{1}_{(t_i,t_{i+1}]^2}(t,s)$$

We thus have:

$$k^{(0)}(\delta, \delta) = \delta^{2}$$
$$\partial_{1}k^{(0)}(\delta, \delta) = \delta$$
$$\partial_{2}k^{(0)}(\delta, \delta) = \delta$$
$$\partial_{1}\partial_{2}k^{(0)}(\delta, \delta) = 1$$

So we have:

$$\begin{split} \tilde{A}_{\delta}V_{0}\tilde{A}_{\delta}^{*} &= M\Lambda M + \delta M\Lambda A + \delta A\Lambda M + \delta^{2}A\Lambda A \\ &= M\Lambda(M + \delta A) + \delta A\Lambda(M + \delta A) \\ &= (M\Lambda + \delta A\Lambda)(M + \delta A) \\ &= (M + \delta A)\Lambda(M + \delta A) = Q\Lambda Q \end{split}$$

where we have defined $Q := M + \delta A$. To make progress we now simplify equation (16) using the derivatives noting the following:

$$k^{(0)}(s,\delta) = s\delta \mathbb{1}_{(0,\delta]}(s)$$
$$\partial_2 k^{(0)}(s,\delta) = s\mathbb{1}_{(0,\delta]}(s)$$

We thus have:

$$(V_0 \tilde{A}_{\delta}^* \boldsymbol{v})(x, s) = s \mathbb{1}_{(0, \delta]}(s) \Phi(x)^* \Lambda M \boldsymbol{v} + s \delta \mathbb{1}_{(0, \delta]}(s) \Phi(x)^* \Lambda A \boldsymbol{v}$$

$$= s \mathbb{1}_{(0, \delta]}(s) \Phi(x)^* \Lambda (M + \delta A) \mathbf{v}$$

$$= s \mathbb{1}_{(0, \delta]}(s) \Phi(x)^* \Lambda Q \mathbf{v}$$
(18)

We can now make progress with the mean update equation. We first work out the following term using (18):

$$\begin{split} (V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} F^{(1)})(x,s) &= s \mathbb{1}_{(0,\delta]}(s) \Phi(x)^* \Lambda Q (Q \Lambda Q)^{-1} F^{(1)} \\ &= s \mathbb{1}_{(0,\delta]}(s) \Phi(x)^* \Lambda Q Q^{-1} \Lambda^{-1} Q^{-1} F^{(1)} \\ &= s \mathbb{1}_{(0,\delta]}(s) \Phi(x)^* Q^{-1} F^{(1)} \end{split}$$

For the other term involving m_0 in the mean update equation we must first work out $\tilde{A}_{\delta}m_0 = \mathcal{I}_{\delta}\mathcal{L}m_0$. To do this we compute:

$$(\mathcal{L}m_0)(x,t) = \partial_t m_0 - \nabla \cdot (a(x)\nabla m_0(x,t))$$

$$= -\nabla (a(x)\nabla \Phi(x)^* \gamma)$$

$$= -\sum_{j=1}^J \gamma_j \nabla \cdot (a(x)\nabla \phi_j(x))$$

Thus, the *i-th* entry of $\tilde{A}_{\delta}m_0$ can be computed as:

$$(\tilde{A}_{\delta}m_0)_i = I_i(\delta)\mathcal{L}m_0$$

$$= \int_{\Omega} \phi_i(x) \left(-\sum_{j=1}^J \gamma_j \nabla \cdot (a(x)\nabla \phi_j(x)) \right) dx$$

$$= \sum_{j=1}^J \gamma_j A_{ij} = (A\gamma)_i$$

So $\tilde{A}_{\delta}m_0 = A\gamma$. We can now compute the second term involving m_0 as follows (again using (18)):

$$\begin{split} ((V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} \tilde{A}_{\delta}) m_0)(x,s) &= (V_0 \tilde{A}_{\delta}^* (Q \Lambda Q)^{-1} A \gamma)(x,s) \\ &= s \mathbb{1}_{(0,\delta]} (s) \Phi(x)^* \Lambda Q Q^{-1} \Lambda^{-1} Q^{-1} A \gamma \\ &= s \mathbb{1}_{(0,\delta]} (s) \Phi(x)^* Q^{-1} A \gamma \end{split}$$

Thus we can compute:

$$(1 - (V_0 \tilde{A}_{\delta}^* (\tilde{A}_{\delta} V_0 \tilde{A}_{\delta}^*)^{-1} \tilde{A}_{\delta}) m_0)(x, s) = \Phi(x)^* \gamma - s \mathbb{1}_{(0, \delta]}(s) \Phi(x)^* Q^{-1} A \gamma$$

$$= \Phi(x)^* Q^{-1} [Q - s \mathbb{1}_{(0, \delta]}(s) A] \gamma$$

$$= \Phi(x)^* Q^{-1} [M + \delta A - s \mathbb{1}_{(0, \delta]}(s) A] \gamma$$

$$= \Phi(x)^* Q^{-1} [M + (\delta - s \mathbb{1}_{(0, \delta]}(s)) A] \gamma$$

Putting this all together we obtain:

$$m_1(x,s) = \Phi(x)^* Q^{-1} [M + (\delta - s \mathbb{1}_{(0,\delta]}(s)) A] \gamma + s \mathbb{1}_{(0,\delta]}(s) \Phi(x)^* Q^{-1} F^{(1)}$$
(19)

Thus we see that performing this mean update and then evaluating at the time $t_1 = \delta$ we obtain the following:

$$m_1(x, t_1) = m_1(x, \delta) = \Phi(x)^* \left(Q^{-1} M \gamma + \delta Q^{-1} F^{(1)} \right)$$
 (20)

Note that this is the same as the update equation for the coefficients in the backward-Euler Galerkin method:

$$\gamma \longmapsto (M + \delta A)^{-1} M \gamma + \delta (M + \delta A)^{-1} F^{(1)} \tag{21}$$

!!! Before proceeding to compute the covariance V_1 we first note that m_1 can be expressed as follows:

$$m_1(x,s) = c(x,s) + \Psi^* Q^{-1} \mathcal{I}_{\delta} f$$

i.e. $m_1 = Lf + c$

where $c(x,s) = \Phi(x)^*Q^{-1}\left[M + (\delta - s\mathbb{1}_{(0,\delta]}(s))A\right]\gamma$, Ψ^* is the following function from $\mathbb{R}^J \to L^2(\Omega \times [0,T])$ given by $(\Psi^*\mathbf{v})(x,s) := s\mathbb{1}_{(0,\delta]}(s)\Phi(x)^*\mathbf{v}$ and $L := \Psi^*Q^{-1}\mathcal{I}_{\delta}$. This observation will prove useful later when we marginalise over the noise term. !!!

We can now move on to computing the covariance V_1 . We start by computing $\tilde{A}_{\delta}V_0$. Computing this involves determining how $I_i(\delta)\mathcal{L}V_0$ acts on functions g for $j=1,\ldots,J$. We have:

$$I_j(\delta)\mathcal{L}V_0g = \int_{\Omega} \phi_j(x)(\mathcal{L}V_0g)(x,\delta)dx$$

Now recalling that $V_0g(x,\delta) = \int_{\Omega} \int_0^T k_{x\delta,ys}^{(0)} g(y,s) dsdy$ we deduce:

$$(\mathcal{L}V_0g)(x,\delta) = \int_{\Omega} \int_0^T (\mathcal{L}k_{\cdot,\cdot,y,s}^{(0)})(x,\delta)g(y,s)\mathrm{d}s\mathrm{d}y$$

$$= \int_{\Omega} \int_0^T \left(\partial_1 k^{(0)}(\delta,s) \sum_{i=1}^J \lambda_i \phi_i(x)\phi_i(y) - k^{(0)}(\delta,s) \sum_{i=1}^J \lambda_i \nabla_x \cdot (a(x)\nabla_x \phi_i(x))\phi_i(y) \right) g(y,s)\mathrm{d}s\mathrm{d}y$$

We can now perform the integration to obtain:

$$\begin{split} I_{j}(\delta)\mathcal{L}V_{0}g &= \int_{\Omega}\phi_{j}(x)\left(\int_{\Omega}\int_{0}^{T}\left(\partial_{1}k^{(0)}(\delta,s)\sum_{i=1}^{J}\lambda_{i}\phi_{i}(x)\phi_{i}(y) - k^{(0)}(\delta,s)\sum_{i=1}^{J}\lambda_{i}\nabla_{x}\cdot(a(x)\nabla_{x}\phi_{i}(x))\phi_{i}(y)\right)g(y,s)\mathrm{d}s\mathrm{d}y\right)\mathrm{d}x\\ &= \int_{\Omega}\int_{0}^{T}\left(\partial_{1}k^{(0)}(\delta,s)\sum_{i=1}^{J}\lambda_{i}M_{ij}\phi_{i}(y)g(y,s) + k^{(0)}(\delta,s)\sum_{i=1}^{J}\lambda_{i}A_{ij}\phi_{i}(y)g(y,s)\right)\mathrm{d}s\mathrm{d}y\\ &= \sum_{i=1}^{J}\lambda_{i}\int_{\Omega}\int_{0}^{T}(\partial_{1}k^{(0)}(\delta,s)M_{ij} + k^{(0)}(\delta,s)A_{ij})\phi_{i}(y)g(y,s)\mathrm{d}s\mathrm{d}y\\ &= \sum_{i=1}^{J}\lambda_{i}\left[\int_{0}^{T}M_{ij}\partial_{1}k^{(0)}(\delta,s)(I_{i}(s)g)\mathrm{d}s + \int_{0}^{T}A_{ij}k^{(0)}(\delta,s)(I_{i}(s)g)\mathrm{d}s\right]\\ &= \sum_{i=1}^{J}\lambda_{i}\left[M_{ij}\left(\int_{0}^{T}\partial_{1}k^{(0)}(\delta,s)(\mathcal{I}_{s}g)\mathrm{d}s\right)_{i} + A_{ij}\left(\int_{0}^{T}k^{(0)}(\delta,s)(\mathcal{I}_{s}g)\mathrm{d}s\right)_{i}\right]\\ &= \left(M\Lambda\int_{0}^{T}\partial_{1}k^{(0)}(\delta,s)(\mathcal{I}_{s}g)\mathrm{d}s + A\Lambda\int_{0}^{T}k^{(0)}(\delta,s)(\mathcal{I}_{s}g)\mathrm{d}s\right)_{i}. \end{split}$$

Thus we can deduce:

$$\tilde{A}_{\delta}V_{0}g = M\Lambda \int_{0}^{T} \partial_{1}k^{(0)}(\delta, s)(\mathcal{I}_{s}g)ds + A\Lambda \int_{0}^{T} k^{(0)}(\delta, s)(\mathcal{I}_{s}g)ds$$
(22)

We now utilise the specific form of the time kernel:

$$k^{(0)}(\delta, s) = s\delta \mathbb{1}_{(0,\delta]}(s)$$
$$\partial_1 k^{(0)}(\delta, s) = s\mathbb{1}_{(0,\delta]}(s)$$

to further deduce:

$$\tilde{A}_{\delta}V_{0}g = M\Lambda \int_{0}^{T} s\mathbb{1}_{(0,\delta]}(s)(\mathcal{I}_{s}g)ds + A\Lambda \int_{0}^{T} s\delta\mathbb{1}_{(0,\delta]}(s)(\mathcal{I}_{s}g)ds$$
$$= Q\Lambda \nu_{g}^{(0)}$$

where:

$$\nu_g^{(i)} := \int_{t_i}^{t_{i+1}} (s - t_i)(\mathcal{I}_s g) ds \text{ for } i = 0, \dots, N - 1$$
(23)

Having worked this out we can now compute the second term in the formula for V_1 as follows:

$$(V_{0}\tilde{A}_{\delta}^{*}(\tilde{A}_{\delta}V_{0}\tilde{A}_{\delta}^{*})^{-1}\tilde{A}_{\delta}V_{0}g)(x,t) = (V_{0}\tilde{A}_{\delta}^{*}(Q\Lambda Q)^{-1}Q\Lambda \nu_{g}^{(0)})(x,t)$$

$$= t\mathbb{1}_{(0,\delta]}(t)\Phi(x)^{*}\Lambda QQ^{-1}\Lambda^{-1}Q^{-1}Q\Lambda \nu_{g}^{(0)}$$

$$= t\mathbb{1}_{(0,\delta]}(t)\Phi(x)^{*}\Lambda \nu_{g}^{(0)}$$

One can also easily show that the action of V_0 on functions can be rewritten as follows:

$$(V_0 g)(x,t) = \sum_{i=0}^{N-1} (t - t_i) \mathbb{1}_{(t_i, t_{i+1}]}(t) \Phi(x)^* \Lambda \nu_g^{(i)}$$
(24)

and so we can deduce that V_1 is given by:

$$(V_1 g)(x,t) = \sum_{i=1}^{N-1} (t - t_i) \mathbb{1}_{(t_i, t_{i+1}]}(t) \Phi(x)^* \Lambda \nu_g^{(i)}$$

This can be rewritten as follows:

$$(V_1 g)(x,t) = \int_{\Omega} \int_0^T k_{x,t,y,s}^{(1)} g(y,s) ds dy$$
 (25)

where the kernel is now:

$$k_{x,t,y,s}^{(1)} := \sum_{i=1}^{J} \lambda_j \phi_j(x) \phi_j(y) k^{(1)}(t,s)$$
 (26)

with time kernel:

$$k^{(1)}(t,s) := \sum_{i=1}^{N-1} (t - t_i)(s - t_i) \mathbb{1}_{(t_i, t_{i+1}]^2}(t,s)$$
(27)

We can thus see that the new covariance operator is of the same form as the prior covariance; the only difference is the time kernel has changed by simply removing the term corresponding to the first time interval $(t_0, t_1]$. This observation will allow us to very easily move to the next time step $t = t_2 = 2\delta$. To do so we now condition on observing $\mathcal{I}_{2\delta}\mathcal{L}u = \mathcal{I}_{2\delta}f =: F^{(2)}$. As before let $\tilde{A}_{2\delta} = \mathcal{I}_{2\delta}\mathcal{L}$. For the same fixed realisation of f we seek the following conditional distribution:

$$u|\{\tilde{A}_{\delta}u = F^{(1)}, \tilde{A}_{2\delta}u = F^{(2)}, f\} \sim \mathcal{N}(m_2, V_2)$$
 (28)

where we have the update equations:

$$m_2 = m_1 + V_1 \tilde{A}_{2\delta}^* (\tilde{A}_{2\delta} V_1 \tilde{A}_{2\delta}^*)^{-1} (F^{(2)} - \tilde{A}_{2\delta} m_1)$$
(29)

$$V_2 = V_1 - V_1 \tilde{A}_{2\delta}^* (\tilde{A}_{2\delta} V_1 \tilde{A}_{2\delta}^*)^{-1} \tilde{A}_{2\delta} V_1$$
(30)

Since the covariance operator V_1 has the same form as V_0 but with a different time kernel most of the computations from before are the same. We give the results, and note where there are differences. First we have the analogue of (16):

$$(V_1 \tilde{A}_{2\delta}^* \boldsymbol{v})(x,s) = \partial_2 k^{(1)}(s,2\delta) \Phi(x)^* \Lambda M \boldsymbol{v} + k^{(1)}(s,2\delta) \Phi(x)^* \Lambda A \boldsymbol{v}$$
(31)

The analogue of (17) is given by:

$$\tilde{A}_{2\delta}V_1\tilde{A}_{2\delta}^* = \partial_1\partial_2 k^{(1)}(2\delta, 2\delta)M\Lambda M + \partial_1 k^{(1)}(2\delta, 2\delta)M\Lambda A + \partial_2 k^{(1)}(2\delta, 2\delta)A\Lambda M + k^{(1)}(2\delta, 2\delta)A\Lambda A \tag{32}$$

The various derivatives for our new time kernel follow almost exactly the same as before:

$$\partial_1 k^{(1)}(t,s) = \sum_{i=1}^{N-1} (s-t_i) \mathbb{1}_{(t_i,t_{i+1}]^2}(t,s)$$

$$\partial_2 k^{(1)}(t,s) = \sum_{i=1}^{N-1} (t-t_i) \mathbb{1}_{(t_i,t_{i+1}]^2}(t,s)$$

$$\partial_1 \partial_2 k^{(1)}(t,s) = \sum_{i=1}^{N-1} \mathbb{1}_{(t_i,t_{i+1}]^2}(t,s)$$

We thus have:

$$k^{(1)}(\delta, \delta) = \delta^{2}$$

$$\partial_{1}k^{(1)}(\delta, \delta) = \delta$$

$$\partial_{2}k^{(1)}(\delta, \delta) = \delta$$

$$\partial_{1}\partial_{2}k^{(1)}(\delta, \delta) = 1$$

$$k^{(1)}(s, 2\delta) = \delta(s - \delta)\mathbb{1}_{(\delta, 2\delta]}(s)$$

$$\partial_{2}k^{(1)}(s, 2\delta) = (s - \delta)\mathbb{1}_{(\delta, 2\delta]}(s)$$

Using this we can now simplify (31) to:

$$(V_1 \tilde{A}_{2\delta}^* \mathbf{v})(x, s) = (s - \delta) \mathbb{1}_{(\delta, 2\delta]}(s) \Phi(x)^* \Lambda Q \mathbf{v}$$
(33)

and (32) to:

$$\tilde{A}_{2\delta}V_1\tilde{A}_{2\delta}^* = Q\Lambda Q \tag{34}$$

We now need to compute the term $\tilde{A}_{2\delta}m_1$. Doing this will involve first computing $\mathcal{L}m_1$. Rewriting m_1 as follows will help with this:

$$m_1(x,s) = \Phi(x)^* Q^{-1} [M + \delta A] \gamma + s \mathbb{1}_{(0,\delta]}(s) \Phi(x)^* Q^{-1} [F^{(1)} - A \gamma]$$

= $\Phi(x)^* \gamma + s \mathbb{1}_{(0,\delta]}(s) \Phi(x)^* \mathbf{c}^{(1)}$

where $c^{(1)} := Q^{-1} [F^{(1)} - A\gamma]$. We can now compute more easily:

$$(\mathcal{L}m_1)(x,s) = \mathbb{1}_{(0,\delta]}(s)\Phi(x)^* \mathbf{c}^{(1)} - \sum_{i=1}^{J} (\gamma + s\mathbb{1}_{(0,\delta]}(s)\mathbf{c}^{(1)})_j \nabla \cdot (a(x)\nabla\phi_j(x))$$

This allows us to work out the *i-th* entry of $\tilde{A}_{2\delta}m_1$ as follows:

$$(\tilde{A}_{2\delta}m_1)_i = I_i(2\delta)\mathcal{L}m_1$$

$$= \int_{\Omega} \phi_i(x)(\mathcal{L}m_1)(x, 2\delta) dx$$

$$= -\int_{\Omega} \phi_i(x) \sum_{j=1}^J \gamma_j \nabla \cdot (a(x)\nabla \phi_j(x)) dx$$

$$= \sum_{j=1}^J \gamma_j A_{ij}$$

$$= (A\gamma)_i$$

So $\tilde{A}_{2\delta}m_1 = A\gamma$. These results enable us to work out the following:

$$(V_1 \tilde{A}_{2\delta}^* (\tilde{A}_{2\delta} V_1 \tilde{A}_{2\delta}^*)^{-1} F^{(2)})(x,s) = (s-t_1) \mathbb{1}_{(t_1,t_2]}(s) \Phi(x)^* Q^{-1} F^{(2)}$$
$$(V_1 \tilde{A}_{2\delta}^* (\tilde{A}_{2\delta} V_1 \tilde{A}_{2\delta}^*)^{-1} \tilde{A}_{2\delta} m_1)(x,s) = (s-t_1) \mathbb{1}_{(t_1,t_2]}(s) \Phi(x)^* Q^{-1} A \gamma$$

Combining these results yields the new mean:

$$m_{2}(x,s) = m_{1}(x,s) + (s-t_{1})\mathbb{1}_{(t_{1},t_{2}]}\Phi(x)^{*}Q^{-1}\left[F^{(2)} - A\gamma\right]$$

$$= m_{1}(x,s) + (s-t_{1})\mathbb{1}_{(t_{1},t_{2}]}(s)\Phi(x)^{*}\boldsymbol{c}^{(2)}$$

$$= \Phi(x)^{*}\gamma + (s-t_{0})\mathbb{1}_{(t_{0},t_{1}]}(s)\Phi(x)^{*}\boldsymbol{c}^{(1)} + (s-t_{1})\mathbb{1}_{(t_{1},t_{2}]}(s)\Phi(x)^{*}\boldsymbol{c}^{(2)}$$

$$= \Phi(x)^{*}\gamma + \sum_{i=1}^{2}(s-t_{i-1})\mathbb{1}_{(t_{i-1},t_{i}]}(s)\Phi(x)^{*}\boldsymbol{c}^{(i)}$$

Evaluating this new mean at time $t = t_2 = 2\delta$ we obtain:

$$m_{2}(x, t_{2}) = m_{2}(x, 2\delta) = \Phi(x)^{*} \gamma + (2\delta - \delta)\Phi(x)^{*} c^{(2)}$$

$$= \Phi(x)^{*} \gamma + \delta \Phi(x)^{*} Q^{-1} \left[F^{(2)} - A \gamma \right]$$

$$= \Phi(x)^{*} Q^{-1} \left[Q \gamma + \delta F^{(2)} - \delta A \gamma \right]$$

$$= \Phi(x)^{*} Q^{-1} \left[(Q - \delta A) \gamma + \delta F^{(2)} \right]$$

$$= \Phi(x)^{*} Q^{-1} \left[M \gamma + \delta F^{(2)} \right]$$

We now move onto computing V_2 . Since V_1 is of the same form as V_0 with a different time kernel, the computations for V_1 can be reused almost identically to compute V_2 . We have:

$$\begin{split} \tilde{A}_{2\delta}V_0g &= M\Lambda \int_0^T \partial_1 k^{(1)}(2\delta,s)(\mathcal{I}_s g)\mathrm{d}s + A\Lambda \int_0^T k^{(1)}(2\delta,s)(\mathcal{I}_s g)\mathrm{d}s \\ &= M\Lambda \int_0^T (s-t_1)\mathbb{1}_{(t_1,t_2]}(s)(\mathcal{I}_s g)\mathrm{d}s + \delta A\Lambda \int_0^T (s-t_1)\mathbb{1}_{(t_1,t_2]}(s)(\mathcal{I}_s g)\mathrm{d}s \\ &= M\Lambda \int_{t_1}^{t_2} (s-t_1)(\mathcal{I}_s g)\mathrm{d}s + \delta A\Lambda \int_{t_1}^{t_2} (s-t_1)(\mathcal{I}_s g)\mathrm{d}s \\ &= M\Lambda \boldsymbol{\nu}_g^{(1)} + \delta A\Lambda \boldsymbol{\nu}_g^{(1)} \\ &= Q\Lambda \boldsymbol{\nu}_g^{(1)} \end{split}$$

Having worked this out we can now compute the second term in the formula for V_2 as follows:

$$\begin{split} (V_1 \tilde{A}_{2\delta}^* (\tilde{A}_{2\delta} V_1 \tilde{A}_{2\delta}^*)^{-1} \tilde{A}_{2\delta} V_1 g)(x,t) &= (V_1 \tilde{A}_{2\delta}^* (Q \Lambda Q)^{-1} Q \Lambda \boldsymbol{\nu}_g^{(1)})(x,t) \\ &= (t-t_1) \mathbb{1}_{(t_1,t_2]} (t) \Phi(x)^* \Lambda Q Q^{-1} \Lambda^{-1} Q^{-1} Q \Lambda \boldsymbol{\nu}_g^{(1)} \\ &= (t-t_1) \mathbb{1}_{(t_1,t_2]} (t) \Phi(x)^* \Lambda \boldsymbol{\nu}_g^{(1)} \end{split}$$

Thus, we can conclude that V_2 acts on functions g as follows:

$$(V_2 g)(x,t) = \sum_{i=2}^{N-1} (t - t_i) \mathbb{1}_{(t_i, t_{i+1}]}(t) \Phi(x)^* \Lambda \nu_g^{(i)} = \int_{\Omega} \int_0^T k_{x,t,y,s}^{(2)} g(y, s) ds dy$$
 (35)

where the kernel is now:

$$k_{x,t,y,s}^{(2)} := \sum_{i=2}^{J} \lambda_j \phi_j(x) \phi_j(y) k^{(2)}(t,s)$$
(36)

with time kernel:

$$k^{(2)}(t,s) := \sum_{i=2}^{N-1} (t - t_i)(s - t_i) \mathbb{1}_{(t_i, t_{i+1}]^2}(t,s)$$
(37)

We again can see that the form of the covariance operator remains the same as the previous one. We can now hypothesize a formula for conditioning up to time t_p for $p \in \{1, ..., N\}$:

$$u|\{\tilde{A}_{t_1}u = F^{(1)}, \dots, \tilde{A}_{t_p}u = F^{(p)}, f\} \sim \mathcal{N}(m_p, V_p)$$
 (38)

where $F^{(p)} := \mathcal{I}_{t_p} f = \mathcal{I}_{p\delta} f$ and where the mean and covariance are given by:

$$m_p(x,t) := \Phi(x)^* \gamma + \sum_{i=1}^p (t - t_{i-1}) \mathbb{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* \boldsymbol{c}^{(i)}$$
(39)

$$(V_p g)(x,t) := \int_{\Omega} \int_0^T k_{x,t,y,s}^{(p)} g(y,s) ds dy$$
(40)

$$\boldsymbol{c}^{(i)} := Q^{-1} \left[F^{(i)} - A \boldsymbol{\gamma} \right] \tag{41}$$

$$k_{x,t,y,s}^{(p)} := \sum_{i=p}^{N-1} \lambda_j \phi_j(x) \phi_j(y) k^{(p)}(t,s)$$
(42)

$$k^{(p)}(t,s) := \sum_{i=p}^{N-1} (t - t_i)(s - t_i) \mathbb{1}_{(t_i, t_{i+1}]^2}(t,s)$$
(43)

This can probably be shown by induction. The above implies that conditioning at all N time points will yield the following degenerate Gaussian distribution:

$$u|\{\tilde{A}_{t_1}u = F^{(1)}, \dots, \tilde{A}_{t_N}u = F^{(N)}, f\} \sim \mathcal{N}(m_N, 0) = \delta_{m_N}$$
 (44)

since $V_N = 0$. This is a Dirac point mass located at the function:

$$m_N(x,t) = \Phi(x)^* \gamma + \sum_{i=1}^N (t - t_{i-1}) \mathbb{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* \mathbf{c}^{(i)}$$
(45)

We now want to perform the marginalisation over the RHS noise term f. In order to do this it will help to rewrite m_N as follows:

$$\begin{split} m_N(x,t) &= \Phi(x)^* \gamma + \sum_{i=1}^N (t-t_{i-1}) \mathbbm{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* Q^{-1} \left[F^{(i)} - A \gamma \right] \\ &= \Phi(x)^* \gamma - \sum_{i=1}^N (t-t_{i-1}) \mathbbm{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* Q^{-1} A \gamma + \sum_{i=1}^N (t-t_{i-1}) \mathbbm{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* Q^{-1} \mathcal{I}_{t_i} f \\ &= \Phi(x)^* Q^{-1} \left[Q - \sum_{i=1}^N (t-t_{i-1}) \mathbbm{1}_{(t_{i-1},t_i]}(t) A \right] \gamma + \left[\sum_{i=1}^N (t-t_{i-1}) \mathbbm{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* Q^{-1} \mathcal{I}_i \right] f \\ &= c(x,t) + (Lf)(x,t) \end{split}$$

i.e. $m_N = Lf + c$ where:

$$c(x,t) := \Phi(x)^* Q^{-1} \left[Q - \sum_{i=1}^{N} (t - t_{i-1}) \mathbb{1}_{(t_{i-1},t_i]}(t) A \right] \gamma$$
(46)

$$L := \sum_{i=1}^{N} L_{(i)} \tag{47}$$

$$L_{(i)} := \Psi_{(i)}^* Q^{-1} \mathcal{I}_{t_i} \tag{48}$$

where the $\{\Psi_{(i)}^*\}_{i=1}^N$ are operators from $\mathbb{R}^J \to L^2(\Omega \times [0,T])$ given by:

$$(\Psi_{(i)}^* \mathbf{v})(x,t) := (t - t_{i-1}) \mathbb{1}_{(t_{i-1},t_i]}(t) \Phi(x)^* \mathbf{v}$$
(49)

It is easy to show that the adjoint of $\Psi_{(i)}^*$ is $\Psi_{(i)}$ which is defined as:

$$\Psi_{(i)}g := \boldsymbol{\nu}_{\boldsymbol{\sigma}}^{(i-1)} \tag{50}$$

Note: see equation (23) for the definition of $\nu_g^{(i)}$.

We will now marginalise over f in order to obtain the averaged conditional distribution. To do this we will need the following Lemma (which we prove below):

Lemma 0.1. Let $f \sim \mathcal{N}(\bar{f}, K)$ where we assume that this Gaussian measure is on a Hilbert space of functions $\mathcal{H}_1 \subset \mathbb{R}^{\mathcal{X}}$. Suppose that for a fixed realisation of f we have

$$y|f \sim \mathcal{N}(Lf + c, V)$$

where L is a bounded linear operator from \mathcal{H}_1 to another Hilbert space $\mathcal{H}_2 \subset \mathbb{R}^{\mathcal{X}}$ (so y lies in \mathcal{H}_2), c is a deterministic function and the covariance opertor V does not depend on f. Then marginalizing over f yields:

$$y \sim \mathcal{N}(L\bar{f} + c, LKL^* + V)$$

as the averaged distribution of y.

Proof: The fact that $y|f \sim \mathcal{N}(Lf + c, V)$ is equivalent to saying that:

$$y = Lf + c + \tilde{y}$$

where $\tilde{y} \sim \mathcal{N}(0, V)$ is independent of f. Thus we have:

$$\begin{pmatrix} f \\ \tilde{y} \end{pmatrix} = \mathcal{N} \left(\begin{pmatrix} \bar{f} \\ 0 \end{pmatrix}, \begin{pmatrix} K & 0 \\ 0 & V \end{pmatrix} \right)$$

Since we can write:

$$y = \begin{pmatrix} L & 1 \end{pmatrix} \begin{pmatrix} f \\ \tilde{y} \end{pmatrix} + c$$

we deduce:

$$y \sim \mathcal{N}\left(L\bar{f} + c, \begin{pmatrix} L & 1 \end{pmatrix} \begin{pmatrix} K & 0 \\ 0 & V \end{pmatrix} \begin{pmatrix} L^* \\ 1 \end{pmatrix} \right) = \mathcal{N}(L\bar{f} + c, LKL^* + V)$$

as required. \blacksquare

We can now perform the marginalisation over f noting that $V_N = 0$ does not depend on f to obtain:

$$\int u |\{\tilde{A}_{t_1} u = F^{(1)}, \dots, \tilde{A}_{t_N} u = F^{(N)}, f\} df \sim \mathcal{N}(L\bar{f} + c, LKL^*)$$
(51)