

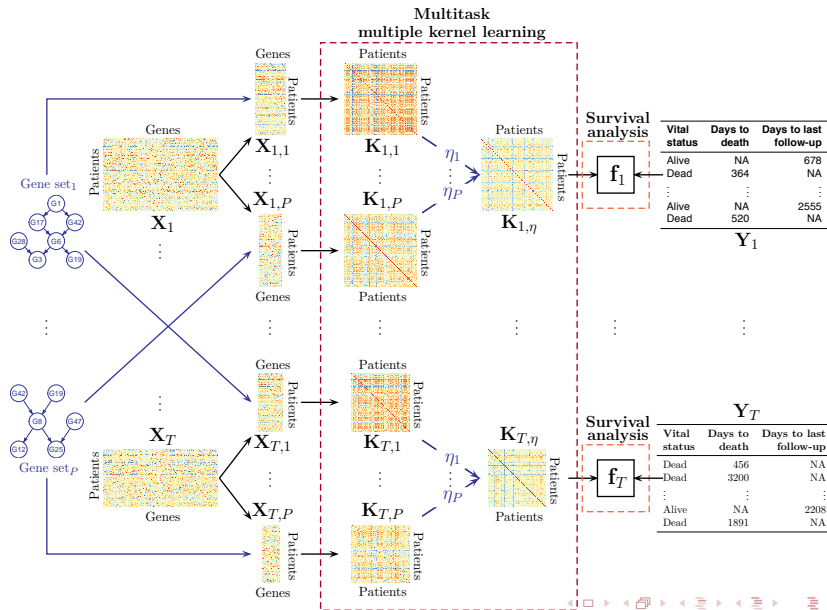
A Multitask Multiple Kernel Learning Algorithm for Survival Analysis with Application to Cancer Biology

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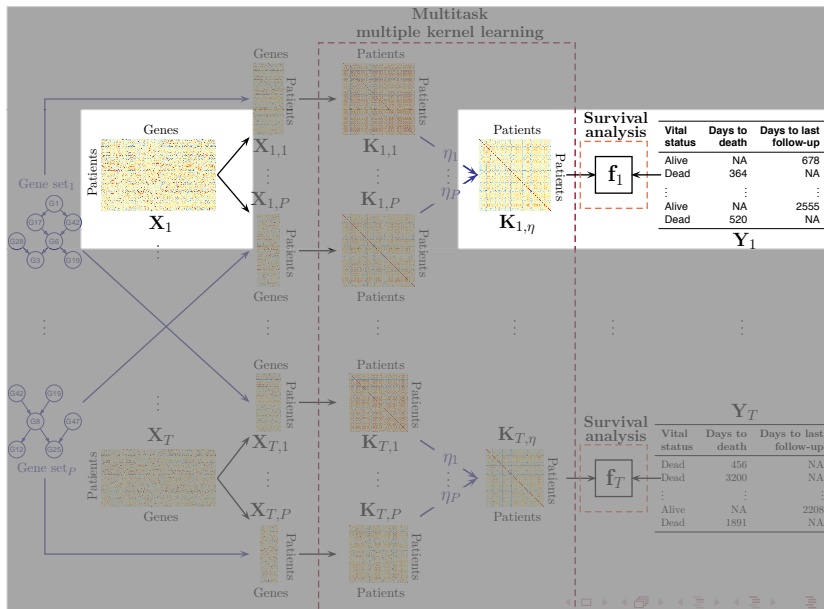
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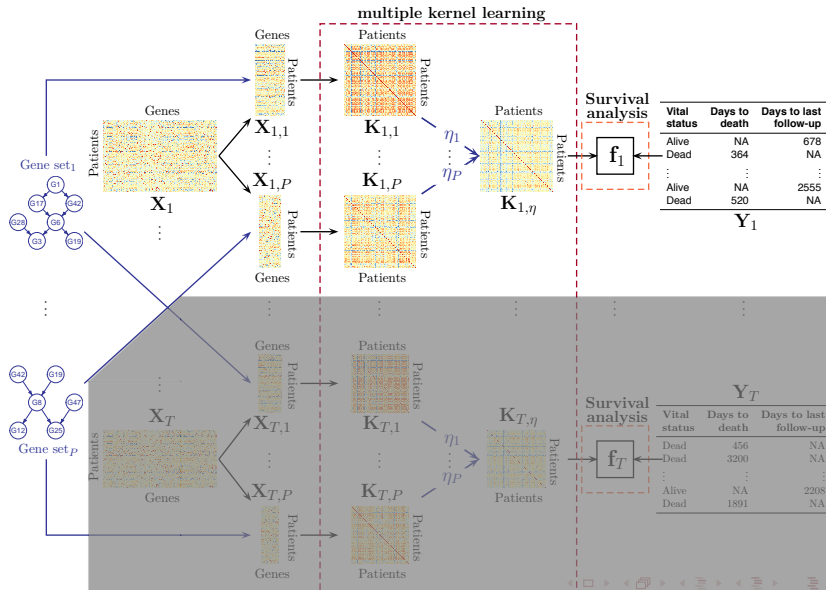
Proposed Approach



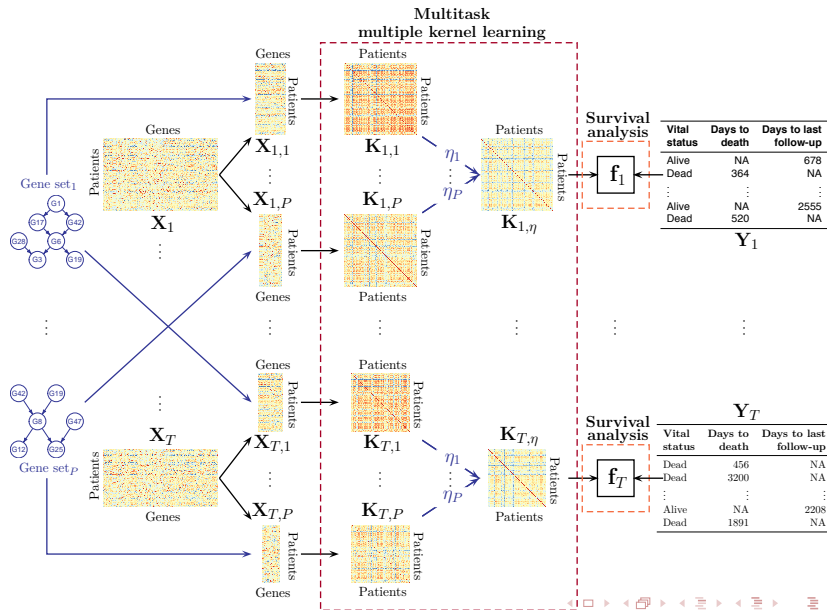
Proposed Approach



Proposed Approach



Proposed Approach



Multitask Survival MKL Formulation

$$\text{minimize } \sum_{t=1}^T \left[\frac{1}{2} \mathbf{w}_t^\top \mathbf{w}_t + C \sum_{i=1}^{N_t} (\xi_{ti}^+ + (1 - \delta_{ti}) \xi_{ti}^-) \right]$$

with respect to $\mathbf{w}_t \in \mathbb{R}^{D_t}$, $\boldsymbol{\xi}_t^+ \in \mathbb{R}^{N_t}$, $\boldsymbol{\xi}_t^- \in \mathbb{R}^{N_t}$, $\mathbf{b}_t \in \mathbb{R}$

$$\text{subject to } \epsilon + \xi_{ti}^+ \geq y_{ti} - \mathbf{w}_t^\top \mathbf{x}_{ti} - \mathbf{b}_t \quad \forall(t, i)$$

$$\epsilon + \xi_{ti}^- \geq \mathbf{w}_t^\top \mathbf{x}_{ti} + \mathbf{b}_t - y_{ti} \quad \forall(t, i)$$

$$\xi_{ti}^+ \geq 0 \quad \forall(t, i)$$

$$\xi_{ti}^- \geq 0 \quad \forall(t, i)$$

Multitask Survival MKL Formulation

$$\begin{aligned} \text{minimize} \quad & - \sum_{t=1}^T \sum_{i=1}^{N_t} y_{ti} (\alpha_{ti}^+ - \alpha_{ti}^-) + \epsilon \sum_{t=1}^T \sum_{i=1}^{N_t} (\alpha_{ti}^+ + \alpha_{ti}^-) \\ & + \frac{1}{2} \sum_{t=1}^T \sum_{i=1}^{N_t} \sum_{j=1}^{N_t} (\alpha_{ti}^+ - \alpha_{ti}^-) (\alpha_{tj}^+ - \alpha_{tj}^-) \sum_{m=1}^P \eta_m k_m(\mathbf{x}_{ti}, \mathbf{x}_{tj}) \end{aligned}$$

with respect to $\alpha_t^+ \in \mathbb{R}^{N_t}$, $\alpha_t^- \in \mathbb{R}^{N_t}$, $\eta \in \mathbb{R}^P$

$$\text{subject to} \quad \sum_{i=1}^{N_t} (\alpha_{ti}^+ - \alpha_{ti}^-) = 0 \quad \forall t$$

$$C \geq \alpha_{ti}^+ \geq 0 \quad \forall(t, i)$$

$$C(1 - \delta_{ij}) \geq \alpha_{ij}^- \geq 0 \quad \forall(t, i)$$

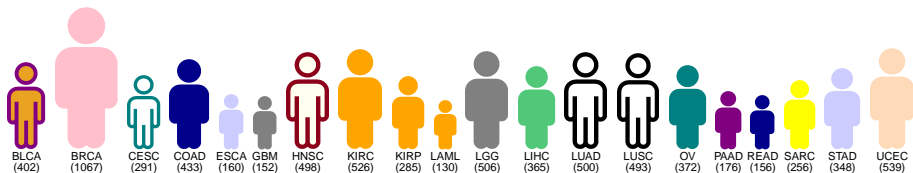
$$\sum_{m=1}^P \eta_m = 1$$

$$\eta_m \geq 0 \quad \forall m$$

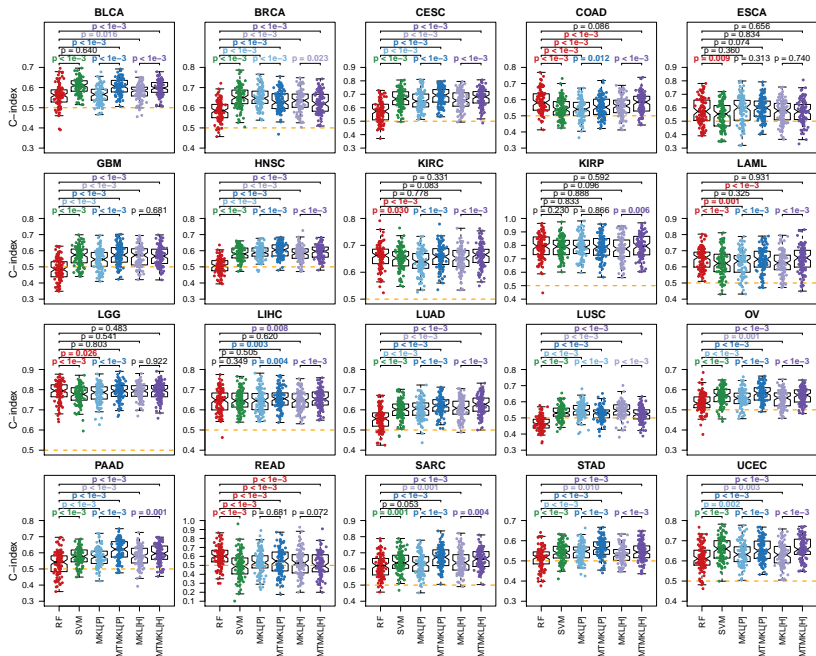
Multitask Survival MKL Formulation

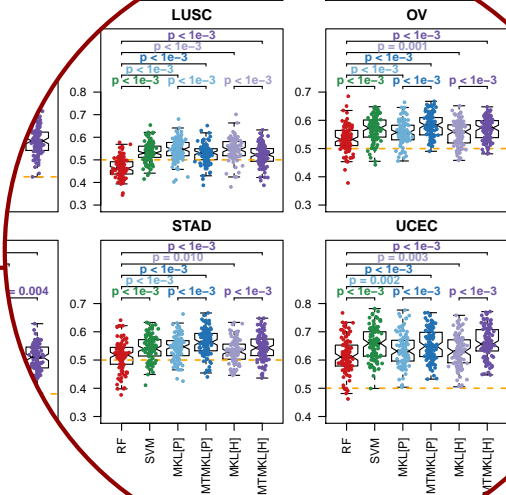
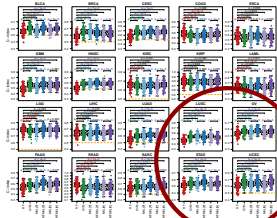
$$\eta_m^{(s+1)} = \frac{\sum_{t=1}^T \eta_m^{(s)} \sqrt{\sum_{i=1}^{N_t} \sum_{j=1}^{N_t} \alpha_{ti}^{(s)} \alpha_{tj}^{(s)} k_m(\mathbf{x}_{ti}, \mathbf{x}_{tj})}}{\sum_{t=1}^T \sum_{o=1}^P \eta_o^{(s)} \sqrt{\sum_{i=1}^{N_t} \sum_{j=1}^{N_t} \alpha_{ti}^{(s)} \alpha_{tj}^{(s)} k_o(\mathbf{x}_{ti}, \mathbf{x}_{tj})}} \quad \forall m$$

Data Sets



- 20 cancer data sets from TCGA database
Gene expression profiles and **survival characteristics**
- Hallmark Gene Set [1] & PID Pathway [2] Collections





Summary

- A multitask multiple kernel learning algorithm
- Integration of different data sets
- Performing survival prediction and knowledge extraction conjointly
- Understanding the underlying mechanisms of cancer

References

- [1] A. Liberzon, C. Birger, H. Thorvaldsdottir, M. Ghandi, J. P. Mesirov, and P. Tamayo. The Molecular Signatures Database (MSigDB) hallmark gene set collection. *Cell Syst.*, 1:417–425, 2015.
- [2] C. F. Schaefer, K. Anthony, S. Krupa, J. Buchoff, M. Day, et al. PID: The Pathway Interaction Database. *Nucleic Acids Res.*, 37:D674–D679, 2009.

Thank you

You can reach R implementations of our work at
<https://github.com/mehmetgonen/path2msurv>.

Please visit our poster if you have any questions.

Room: Pacific Ballroom #242

Date: Wed Jun 12th 06:30 - 09:00 PM

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