

MSK-CHORD: Clinical Outcomes

MSK CHORD Study

- For this class, you will be conducting two projects using data from the MSK CHORD study.
- The data is available in the cBioPortal data repository for cancer genomics.
- You can download the MSK-CHORD data here:
https://www.cbioportal.org/study/summary?id=msk_chord_2024

Common Survival Outcomes

- When defining survival outcomes, an important thing to choose is the “starting time”.
- For this type of data, three common ways to define the starting time are
 - **Time of Diagnosis**
 - **Time of First Metastasis**
 - **Time of First Treatment (for a specific treatment of interest)**

Time from Diagnosis

- Time from cancer diagnosis is a very common starting point.

OS from Diagnosis

- Overall Survival (OS) is probably the most common or important survival endpoint used.
- The most meaningful endpoint in many cases.
- A more “clear” endpoint.
 - Less dependent on details of how the study was conducted.
 - OS results are often more consistent across studies.

OS from Diagnosis

- Let's try to quantify OS from diagnosis by using the `data_timeline_diagnosis.txt` and `data_clinical_patient.txt` files

```
1 library(dplyr)
2 DiagnosisFull <- read.delim2("~/Downloads/msk_chord_2024/data_timeline_diagnosis.txt")
3 DiagnosisFull <- DiagnosisFull %>%
4     arrange(PATIENT_ID, START_DATE)
5
6 ## For people with multiple diagnoses, only keep rows of first diagnosis
7 DiagnosisFirst <- DiagnosisFull[!duplicated(DiagnosisFull$PATIENT_ID),]
```

- Now, load `data_clinical_patient.txt` and merge the two files

```
1 ClinOutcomes <- read.delim("~/Downloads/msk_chord_2024/data_clinical_patient.txt",
2                               comment.char="#")
3 ClinDiag <- inner_join(DiagnosisFirst, ClinOutcomes, by="PATIENT_ID")
```

- Keep only the variables we need for now:

```
1 ClinDiag <- ClinDiag %>%
2     select(PATIENT_ID, START_DATE, OS_MONTHS, OS_STATUS)
```

OS from Diagnosis

```
1 head(ClinDiag)
```

	PATIENT_ID	START_DATE	OS_MONTHS	OS_STATUS
1	P-0000012	-9641	118.45466	0:LIVING
2	P-0000015	-2559	13.90683	1:DECEASED
3	P-0000036	-315	115.46289	0:LIVING
4	P-0000041	-3334	13.61094	1:DECEASED
5	P-0000057	-1036	29.62189	1:DECEASED
6	P-0000058	-205	60.75610	1:DECEASED

- To quantify **OS from Diagnosis**, we must create a new variable which records **months of follow up from time of diagnosis**

```
1 ## Remember START_DATE measures things in days
2 ClinDiag$OS_FROM_DIAG <- (ClinDiag$OS_MONTHS*30.4375 - ClinDiag$START_DATE)/30.4375
3 head(ClinDiag)
```

	PATIENT_ID	START_DATE	OS_MONTHS	OS_STATUS	OS_FROM_DIAG
1	P-0000012	-9641	118.45466	0:LIVING	435.20210
2	P-0000015	-2559	13.90683	1:DECEASED	97.98076
3	P-0000036	-315	115.46289	0:LIVING	125.81196
4	P-0000041	-3334	13.61094	1:DECEASED	123.14688
5	P-0000057	-1036	29.62189	1:DECEASED	63.65885
6	P-0000058	-205	60.75610	1:DECEASED	67.49121

OS from Diagnosis

- I use the conversion **1 month = 30.4375 days**.
 - I'm not sure this conversion matches how they constructed months in MSK-CHORD, but it should be pretty close.
- To estimate **OS survival curves**, you need to convert the variable `OS_STATUS` to either numeric or logical variable

```
1 ClinDiag$OS_STATUS_NUM <- ifelse(ClinDiag$OS_STATUS=="1:DECEASED", 1, 0)
```

OS from Diagnosis

- To estimate a **survival curve** in R, you can use the `survfit` function from the `survival` package in R

```
1 library(survival)
2
3 ## Use the format: Surv(time, status) when using survfit
4 os_all_fit <- survfit(Surv(OS_FROM_DIAG, OS_STATUS_NUM) ~ 1, data=ClinDiag)
5 os_all_fit
```

Call: `survfit(formula = Surv(OS_FROM_DIAG, OS_STATUS_NUM) ~ 1, data = ClinDiag)`

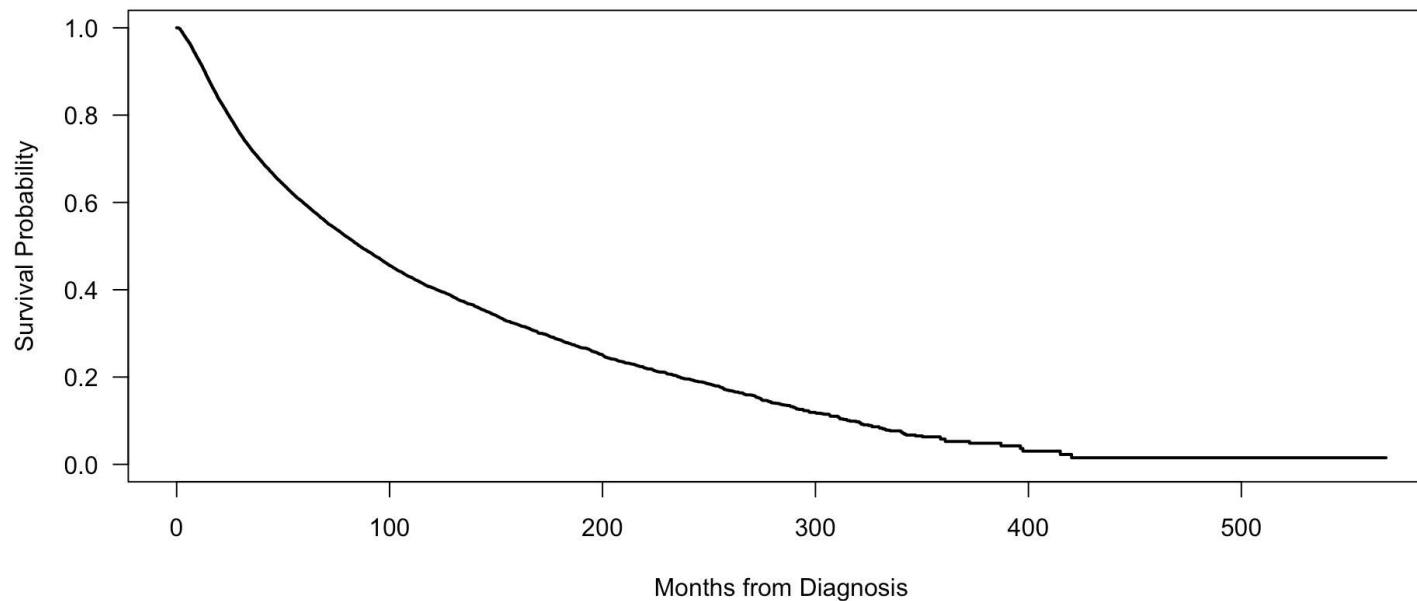
	n	events	median	0.95LCL	0.95UCL
[1,]	24940	11290	86	83.7	88.5

- Grouping all patients together, median survival was **86 months** from diagnosis, with a 95% confidence interval of (83.7 - 88.5).

OS from Diagnosis

- You can plot the full OS survival curve by just plotting the object returned by `survfit`

```
1 plot(os_all_fit, xlab="Months from Diagnosis", ylab="Survival Probability",
2       las=1, lwd=2, conf.int=FALSE)
```



OS from Diagnosis (by Cancer Type)

- To estimate survival for **different subgroups**, you can use the `Surv(time, status) ~ subgrp` syntax.
- I merged `ClinDiag` with the data from `data_clinical_sample.txt` to get cancer type information.
- To estimate OS across types, we can use `survfit` in the following way:

```
1 os_by_type <- survfit(Surv(OS_FROM_DIAG, OS_STATUS_NUM) ~ CANCER_TYPE, data=ClinDiagSamp)
```

OS from Diagnosis (by Cancer Type)

```
1 os_by_type
```

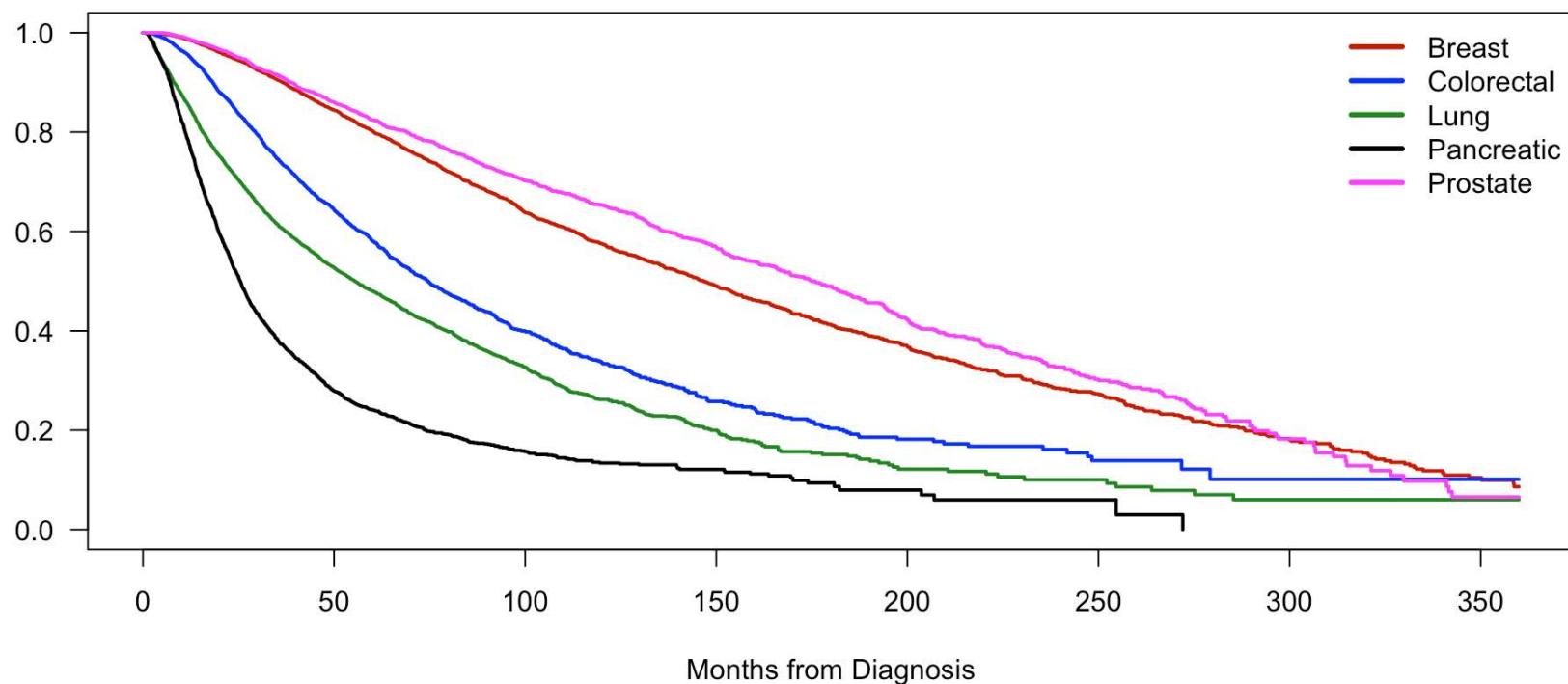
```
Call: survfit(formula = Surv(OS_FROM_DIAG, OS_STATUS_NUM) ~ CANCER_TYPE,  
             data = ClinDiagSamp)
```

	n	events	median	0.95LCL	0.95UCL
CANCER_TYPE=Breast Cancer	5354	1983	146.5	139.8	154.0
CANCER_TYPE=Colorectal Cancer	5527	2212	74.5	70.9	78.5
CANCER_TYPE=Non-Small Cell Lung Cancer	7760	3949	55.4	53.1	58.2
CANCER_TYPE=Pancreatic Cancer	3095	2110	25.5	24.4	26.6
CANCER_TYPE=Prostate Cancer	3204	1036	174.9	165.8	185.5

- There is considerable heterogeneity in median OS across cancer types:
 - Pancreatic Cancer: 25.6 months
 - Prostate Cancer: 174.6 months

OS from Diagnosis (by Cancer Type)

```
1 plot(os_by_type, xlab="Months from Diagnosis", xlim=c(0, 360), las=1,
2      col=c("red3", "blue", "forestgreen", "black", "magenta"), lwd=2)
3 legend("topright", legend=c("Breast", "Colorectal", "Lung", "Pancreatic", "Prostate"),
4        col=c("red3", "blue", "forestgreen", "black", "magenta"), lwd=3, bty='n')
```



Time of Metastasis in MSK-CHORD

- Metastasis probably represents the major dividing point in cancer.
- Metastatic vs. non-metastatic (local) represents a major change in
 - Prognosis
 - Which treatment strategies are appropriate
 - How treatments are evaluated
 - Can represent major change in tumor biology.
- For many patients, cancer is already metastatic when cancer is first diagnosed (de novo metastatic disease).
- For many other patients, cancer is not metastatic but later metastasizes.

Time of Metastasis in MSK-CHORD

- Expected survival can be very different in metastatic vs. non-metastatic patients.
- Grouping these patients together combines **two quite different subgroups**.
- One way to make outcomes more comparable is to eliminate non-metastatic patients.
 - That is, those who were never metastatic for any time.
- Measure survival from **time of metastasis** for the remaining patients.

Time of Metastasis in MSK-CHORD

- There is not a single, structured variable in MSK-CHORD that represents a **clinician-verified time of metastasis**.
- Instead, you need to reconstruct it from the [`data_timeline_tumor_sites.txt`](#) file.

```
1 TumorSites <- read.delim("~/Downloads/msk_chord_2024/data_timeline_tumor_sites.txt")
2 TumorSites <- TumorSites %>%
3     select(PATIENT_ID, START_DATE, TUMOR_SITE, SOURCE_SPECIFIC) %>%
4     arrange(PATIENT_ID, START_DATE)
5 TumorSites[20:26,] ## Look at Patient P-0000015
```

	PATIENT_ID	START_DATE	TUMOR_SITE	SOURCE_SPECIFIC
20	P-0000015	-90	Bone	CT
21	P-0000015	-90	Other	CT
22	P-0000015	-22	Bone	CT
23	P-0000015	-22	Liver	CT
24	P-0000015	-22	Other	CT
25	P-0000015	-22	Pleura	CT
26	P-0000015	-12	Bone	PET

Time of Metastasis in MSK-CHORD

- Patient P-0000015 was originally diagnosed with **breast cancer**.

```
1 TumorSites[20:26,] ## Look at Patient P-0000015
```

	PATIENT_ID	START_DATE	TUMOR_SITE	SOURCE_SPECIFIC
20	P-0000015	-90	Bone	CT
21	P-0000015	-90	Other	CT
22	P-0000015	-22	Bone	CT
23	P-0000015	-22	Liver	CT
24	P-0000015	-22	Other	CT
25	P-0000015	-22	Pleura	CT
26	P-0000015	-12	Bone	PET

- She had a scan at time -90, which indicates tumor presence in **Bone** and an **Other** site.
- If you check the file [data_timeline_diagnosis.txt](#), this patient was diagnosed with **Stage 1 Breast Cancer** at time -2559.
- This indicates the time of first metastasis is at time -90.
- This is $2559 - 90 = 2469$ days after first diagnosis.

Time of Metastasis in MSK-CHORD

- Basically, for those who **did not** have metastasis at time of diagnosis, you can find first date of metastasis by looking at
 - The `START_DATE` at which they had a scan which indicates tumor presence in some non-primary site.
 - Use `data_timeline_tumor_sites.txt` to find this.

Metastasis at time of diagnosis in MSK-CHORD

```
1 head(DiagnosisFirst)
```

	PATIENT_ID	START_DATE	STAGE_CDM_DERIVED
1	P-0000012	-9641	Stage 1-3
3	P-0000015	-2559	Stage 1-3
4	P-0000036	-315	Stage 4
5	P-0000041	-3334	Stage 1-3
6	P-0000057	-1036	Stage 4
7	P-0000058	-205	Stage 4

SUMMARY

1	N/A
3	Localized
4	Distant
5	Localized
6	Distant
7	Distant

- Stage and Diagnosis Summary indicate if metastasis was present at time of diagnosis.

Metastasis at time of diagnosis in MSK-CHORD

```
1 ## Trim white space to right of the SUMMARY variable to clean it up
2 DiagnosisFirst$SUMMARY <- trimws(DiagnosisFirst$SUMMARY, which="right")
3
4 ## Now, tabulate the summaries
5 table(DiagnosisFirst$SUMMARY)
```

	Distant	Distant metastases/systemic disease	
	2542		7506
In situ		Localized	
	200		5742
	N/A	Regional both 2and 3	
	486		1921
Regional by direct extension		Regional nos	
	1318		18
Regional to lymph nodes		Regional, direct extension	
	2082		745
Regional, extension and nodes		Regional, lymph nodes only	
	951		1248
Unknown/Unstaged		Unstaged unknown	
	93		88

Metastasis at time of diagnosis in MSK-CHORD

- Any summary labeled “Distant” is metastasis at time of diagnosis.

```
1 ## Create a variable which indicates a "Distant" summary at diagnosis
2 DiagnosisFirst$Mets_at_diag <- ifelse(DiagnosisFirst$SUMMARY=="Distant"
3                                     | DiagnosisFirst$SUMMARY=="Distant metastases/systemic disease",
4                                     "Yes", "No")
5 table(DiagnosisFirst$Mets_at_diag)
```

No	Yes
14892	10048

- About 41% of patients had metastasis at first diagnosis.

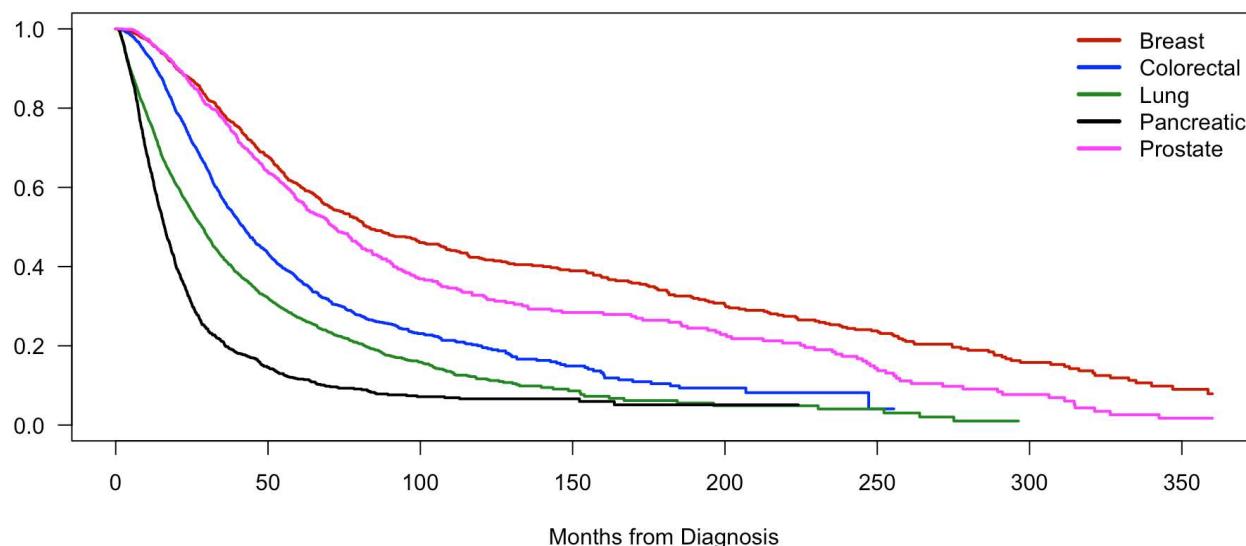
Possible Project Topics

- For prostate cancer, look at time from diagnosis to metastatic disease. What are the genomic characteristics of patients with a long time from initial diagnosis to development of metastasis.

OS from diagnosis (among de novo)

```
Call: survfit(formula = Surv(OS_FROM_DIAG, OS_STATUS_NUM) ~ CANCER_TYPE,  
             data = ClinDiagSampMet)
```

	n	events	median	0.95LCL	0.95UCL
CANCER_TYPE=Breast Cancer	1311	657	83.0	76.8	98.8
CANCER_TYPE=Colorectal Cancer	2580	1529	41.9	39.9	43.9
CANCER_TYPE=Non-Small Cell Lung Cancer	3749	2623	28.4	26.8	29.6
CANCER_TYPE=Pancreatic Cancer	1424	1123	16.0	15.2	16.9
CANCER_TYPE=Prostate Cancer	984	488	71.7	64.9	79.4



Time from Treatment Start

- In some cases, you might want to compare outcomes after a particular type of treatment
- The start time for comparison should be the beginning of treatment
- The treatment initiation data is in the file `data_timeline_treatment.txt`

```
1 Treatment <- read.delim("~/Downloads/msk_chord_2024/data_timeline_treatment.txt")
2 Treatment <- Treatment %>%
3     arrange(PATIENT_ID, START_DATE) %>%
4     select(PATIENT_ID, START_DATE, SUBTYPE, AGENT)
5 head(Treatment)
```

	PATIENT_ID	START_DATE	SUBTYPE	AGENT
1	P-0000012	-5437	Chemo	CYCLOPHOSPHAMIDE
2	P-0000012	-5437	Chemo	FLUOROURACIL
3	P-0000012	-5437	Chemo	METHOTREXATE
4	P-0000012	33	Chemo	CISPLATIN
5	P-0000012	33	Chemo	ETOPOSIDE
6	P-0000012	61	Chemo	CARBOPLATIN

Time from Treatment Start

- You might be interested in a particular class of treatments

```
1 table(Treatment$SUBTYPE)
```

	Biologic	Bone	Treatment	Chemo	Hormone	Immuno
	8816		5635	73036	26670	4030
Investigational			Other	Targeted		
	7284		12	9460		

- For example, **immunotherapies** are a more recent class of treatments

```
1 ImmunoTrt <- subset(Treatment, SUBTYPE=="Immuno")
```

- All of the treatments classified as “Immuno” are really **immune checkpoint inhibitors**

```
1 table(ImmunoTrt$AGENT)
```

ATEZOLIZUMAB	AVELUMAB	CEMIPRIMAB	DURVALUMAB	IPILIMUMAB
389	3	3	254	236
NIVOLUMAB	PEMBROLIZUMAB	TREMELIMUMAB		
886	2258	1		

Time from Treatment Start

- They also classify certain treatments as “**Biologic**”
- Many of these could also be referred to as immunotherapies

```
1 Biologic <- subset(Treatment, SUBTYPE=="Biologic")
2 table(Biologic$AGENT)[1:10]
```

ADO-TRASTUZUMAB	EMTANSINE	AFLIBERCEPT	OPHTHALMIC	ALDESLEUKIN
	352		30	1
	ALEMTUZUMAB		AMIVANTAMAB	BCG
	3		16	103
BELANTAMAB	MAFODOTIN		BEVACIZUMAB	BLINATUMOMAB
	2		3380	1
BORTEZOMIB				
	29			

Time from Treatment Start

- For the immune checkpoint inhibitors (ICI), let's look at what the treatment data looks like:

```
1 head(ImmunoTrt, 8)
```

	PATIENT_ID	START_DATE	SUBTYPE	AGENT
9	P-0000012	1734	Immuno	NIVOLUMAB
171	P-0000113	85	Immuno	DURVALUMAB
355	P-0000165	449	Immuno	NIVOLUMAB
460	P-0000235	381	Immuno	NIVOLUMAB
466	P-0000239	1150	Immuno	NIVOLUMAB
574	P-0000302	929	Immuno	NIVOLUMAB
576	P-0000302	2178	Immuno	IPILIMUMAB
668	P-0000373	977	Immuno	PEMBROLIZUMAB

- Most patients receive an only one ICI
 - But, about 17% (like P-0000302) received more than one ICI

```
1 dim( ImmunoTrt )
```

```
[1] 4030    4
```

```
1 length( unique(ImmunoTrt$PATIENT_ID) )
```

```
[1] 3341
```

Time from Treatment Start

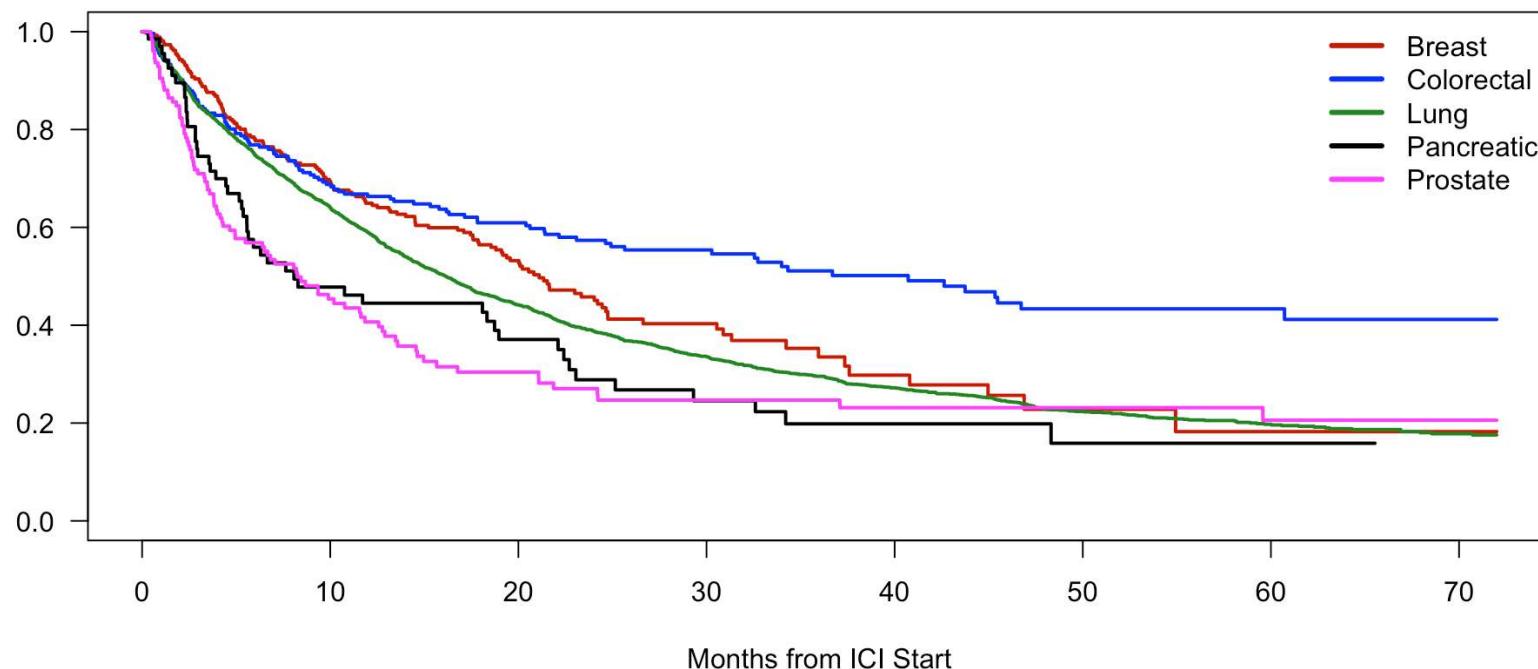
- For patients receiving **more than one ICI**, you would typically look at the **time of first ICI**:

```
1 ImmunoTrtFirst <- ImmunoTrt[!duplicated(ImmunoTrt$PATIENT_ID),]  
2 head(ImmunoTrtFirst, 8)
```

	PATIENT_ID	START_DATE	SUBTYPE	AGENT
9	P-0000012	1734	Immuno	NIVOLUMAB
171	P-0000113	85	Immuno	DURVALUMAB
355	P-0000165	449	Immuno	NIVOLUMAB
460	P-0000235	381	Immuno	NIVOLUMAB
466	P-0000239	1150	Immuno	NIVOLUMAB
574	P-0000302	929	Immuno	NIVOLUMAB
668	P-0000373	977	Immuno	PEMBROLIZUMAB
953	P-0000449	1977	Immuno	ATEZOLIZUMAB

OS from Time of first ICI

- To look at OS from treatment start, you would need to:
 - Subtract time of ICI start from time of last follow up ([OS_MONTHS](#))



Progression-free Survival

- Progression-free survival (PFS) is a common clinical endpoint used in many cancer studies.
- Goal: quantify how long a patient can live **without** having their cancer “progress”.
- Outcome: time to death **OR** time to progression (whichever comes first)
- $Y_i = \min\{T_i, U_i\}$.
 - T_i - time of progression
 - U_i - time of death
- A median PFS of 5 years means that the median of Y_i is 5 years.

Progression-free Survival

- PFS is often measured from the start of a certain treatment.
 - For example, you might report median PFS from the time of starting an ICI.
 - PFS is also commonly measured from time of metastatic disease.

PFS in MSK-CHORD

- In MSK-CHORD, you are really going to be looking at a “**real-world**” measure of **radiographic PFS**.
- “**Real-world**” PFS: This is PFS where scans are not taken at regular, protocol-determined intervals.
 - Scans are taken whenever clinicians/patients think they are appropriate.
- **Radiographic PFS**: Progression is only determined by imaging-based criteria.
- Clinical/Radiographic PFS is used in many cancer studies.
 - Here, progression can include additional factors not captured by imaging.
 - We cannot capture clinical/radiographics PFS using the data in MSK-CHORD.

Radiographic PFS in MSK-CHORD

- To capture radiographic PFS in MSK-CHORD, you will need to use the [data_timeline_progression.txt](#) file.
 - Merging with the “OS from treatment start” data would give you:

	PATIENT_ID	START_DATE_SCAN	START_DATE_TRT	NLP_PROGRESSION_PROBABILITY
156	P-0000239	1008	1150	0.022726787
157	P-0000239	1106	1150	0.044055530
158	P-0000239	1161	1150	0.991534200
159	P-0000239	1204	1150	0.998716100
160	P-0000302	-17	929	0.005758766
161	P-0000302	47	929	0.002286533
162	P-0000302	119	929	0.676131840

	PROCEDURE_TYPE	AGENT	OS_From_Trt	OS_STATUS_NUM
156	CT	NIVOLUMAB	2.162821	1
157	CT	NIVOLUMAB	2.162821	1
158	CT	NIVOLUMAB	2.162821	1
159	CT	NIVOLUMAB	2.162821	1
160	CT	NIVOLUMAB	75.275584	0
161	CT	NIVOLUMAB	75.275584	0
162	CT	NIVOLUMAB	75.275584	0

Radiographic PFS in MSK-CHORD

	PATIENT_ID	START_DATE_SCAN	START_DATE_TRAIT	NLP_PROGRESSION_PROBABILITY
32	P-0000012	1702	1734	0.002305086
33	P-0000012	1862	1734	0.128094140
34	P-0000012	1974	1734	0.015662400
35	P-0000012	2086	1734	0.002995548
36	P-0000012	2269	1734	0.088399045
37	P-0000012	2444	1734	0.111479566
38	P-0000012	2445	1734	0.001614616
39	P-0000012	2627	1734	0.001661092
40	P-0000012	2809	1734	0.003464612
	PROCEDURE_TYPE	AGENT	OS_From_Trt	OS_STATUS_NUM
32	PET	NIVOLUMAB	61.48547	0
33	CT	NIVOLUMAB	61.48547	0
34	CT	NIVOLUMAB	61.48547	0
35	CT	NIVOLUMAB	61.48547	0
36	CT	NIVOLUMAB	61.48547	0
37	MR	NIVOLUMAB	61.48547	0
38	CT	NIVOLUMAB	61.48547	0

Time to Next Treatment

- Time to next treatment (TTNT) is another clinical endpoint that provides a measure of time to progression.

<https://pmc.ncbi.nlm.nih.gov/articles/PMC8085844/>

- Often used as another “**real-world**” measure of progression.
 - Most of the time, it can **reliably** be obtained from health records
- TTNT (usually) captures the period of time under which the disease is stable or improving.
 - Clinicians will not usually recommend the start of a new treatment if the current treatment is working well.
- TTNT can be thought of as a surrogate for **duration of clinical benefit**.

Time to Next Treatment Example

	PATIENT_ID	START_DATE	SUBTYPE	AGENT
364	P-0000175	-1422	Hormone	LEUPROLIDE
365	P-0000175	-1366	Bone Treatment	ZOLEDRONIC ACID
366	P-0000175	-199	Hormone	EXEMESTANE
367	P-0000175	-129	Targeted	EVEROLIMUS
368	P-0000175	420	Hormone	EXEMESTANE
369	P-0000175	537	Chemo	CAPECITABINE
370	P-0000175	707	Hormone	FULVESTRANT
371	P-0000175	707	Targeted	PALBOCICLIB
372	P-0000175	832	Chemo	PACLITAXEL
373	P-0000175	860	Chemo	PACLITAXEL PROTEIN-BOUND
374	P-0000175	1064	Hormone	TAMOXIFEN

- For example, starting from the combination therapy at date 707, the TTNT would be defined as $832 - 707 = 125$ days.